

Stereo and Geometric Stereo

CS635

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Problem Statement



- How to create (realistic) 3D models of existing objects and scenes in the world?
 - Object vs. scene
 - Shape vs. color vs. material properties
 - Automatic vs. manual
 - And many more factors...



<u>http://carlos-hernandez.org/cvpr2010/</u>







Fundamental Approaches

- Manual modeling
 - CAD, Sketchup, 3D Studio Max, MS Paint
- Point Clouds
 - LIDAR, Laser, Kinect
- Photographs
 - "photogrammetry and remote sensing"
 - Single Photograph
 - Stereo Reconstruction (2 photos)
 - Multi-view Reconstruction
 - narrow (video?) or wide baseline



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Definitions



• Camera geometry (=*motion*)

— Given corresponded points on ≥2 views, what are the poses of the cameras?

- Correspondence geometry (=correspondence)
 - Given a point in one view, what are the constraints of its position in another view?
- Scene geometry (=*structure*)
 - Given corresponded points on ≥2 views and the camera poses, what is the 3D location of the points?



Assume that we know P_L corresponds to P_R

Using perspective projection (defined using coordinate system shown)

What is relationship between x_L and X?



Assume that we know P_L corresponds to P_R

Using perspective projection (defined using coordinate system shown)

$$\Rightarrow \quad \frac{x_L}{f} = \frac{X + \frac{b}{2}}{Z} \qquad \qquad \frac{x_R}{f} = \frac{X - \frac{b}{2}}{Z} \qquad \qquad \frac{y_L}{f} = \frac{y_R}{f} = \frac{Y}{Z}$$





 $\Rightarrow d = x_L - x_R$ is the **disparity** between corresponding left and right image points inversely proportional to depth disparity increases with baseline **b**

Stereo: Ray Triangulation scene point optical center optical center image plane

(Ray) Triangulation: compute reconstruction as intersection of two rays

Stereo: Ray Triangulation scene point optical center optical center image plane

Do two lines intersect in 3D? If so, how do you compute their intersection?



Stereo: Ray Triangulation



Equations for the intersection:

$$(p_1 - p_2) \cdot (p_a - p_b) = 0$$

$$(p_3 - p_4) \cdot (p_a - p_b) = 0$$

$$p_b = p_1 + s(p_2 - p_1)$$

$$p_a = p_3 + t(p_4 - p_3)$$

Solve for *s* and *t*, compute *p*:

$$s = \dots$$

$$t = \dots$$

$$p = 0.5(p_a + p_b)$$

Stereo: Vergence field of view of stereo uncertainty of scenepoint one pixel optical axes of the two cameras need not be parallel

Field of view decreases with increase in baseline and vergence
 Accuracy increases with increase in baseline and vergence



 We need to transform "left frame" to "right frame" – includes a rotation and translation:

$$\widetilde{x}_R = R \ \widetilde{x}_L + t_{LR}$$



Camera Geometry

In matrix notation, we can write $\widetilde{x}_R = R \ \widetilde{x}_L + t_{LR}$ as:

$$\widetilde{x}_{L} = \begin{bmatrix} x_{L} \\ y_{L} \\ z_{L} \end{bmatrix} \quad \widetilde{x}_{R} = \begin{bmatrix} x_{R} \\ y_{R} \\ z_{R} \end{bmatrix} \quad R = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} \quad t_{LR} = \begin{bmatrix} r_{14} \\ r_{24} \\ r_{34} \end{bmatrix}$$



Camera Geometry

In matrix notation, we can write $\widetilde{x}_R = R \ \widetilde{x}_L + t_{LR}$ as:

$$r_{11} x_L + r_{12} y_L + r_{13} z_L + r_{14} = x_R$$

$$r_{21} x_L + r_{22} y_L + r_{23} z_L + r_{24} = y_R$$

$$r_{31} x_L + r_{32} y_L + r_{33} z_L + r_{34} = z_R$$

Camera Geometry: Orthonormality Constraints



(a) Rows of R are perpendicular vectors

$$r_{11} r_{21} + r_{12} r_{22} + r_{13} r_{23} = 0$$

$$r_{21} r_{31} + r_{22} r_{32} + r_{23} r_{33} = 0$$

$$r_{11} r_{31} + r_{12} r_{32} + r_{13} r_{33} = 0$$

(b) Each row of R is a unit vector

$$r_{11}^{2} + r_{12}^{2} + r_{13}^{2} = 1$$

$$r_{21}^{2} + r_{22}^{2} + r_{23}^{2} = 1$$

$$r_{31}^{2} + r_{32}^{2} + r_{33}^{2} = 1$$

NOTE: Constraints are NON-LINEAR!



Camera Geometry: Problem Definition

scene



Problem:

Given $\widetilde{x}_L \quad \widetilde{x}_R$'s

Find R $t_{LR} \implies (r_{11}, r_{12}, ..., r_{34})$ subject to (nonlinear) constraints



Problem: same image coords can be generated by doubling $\widetilde{x}_L \ \widetilde{x}_R \ \widetilde{t}_{LR}$ thus, we can find \widetilde{t}_{LR} only up to a scale factor!

Partial

Solution: fix scale by using constraint: $\tilde{t}_{LR} \cdot \tilde{t}_{LR} = 1$ (1 additional equation)





Each scene point gives 3 equations:

 $r_{11} x_L + r_{12} y_L + r_{13} z_L + r_{14} = x_R$ $r_{21} x_L + r_{22} y_L + r_{23} z_L + r_{24} = y_R$ $r_{31} x_L + r_{32} y_L + r_{33} z_L + r_{34} = z_R$

and 6+1 additional equations from orthonormality of rotation matrix constraints and scale constraint.

Thus, for *n* scene points, we have (3n + 6 + 1) equations and 12 unknowns

What is the minimum value for *n*?

Camera Geometry: Solving an Over-determined System



Generally, more than 3 points are used to find the 12 unknowns

Formulate error for scene point *i* as:

$$e_i = (R \ \widetilde{x}_L + t_{LR}) - \widetilde{x}_R$$

Find $R \& t_{LR}$ that minimize: $E = \sum_{i=1}^{N} |e_i|^2 + [\lambda_1 (R^T R - I) + \lambda_2 (t_{LR} \cdot t_{LR} - 1)]$

Camera Geometry: A Linear Estimation



Assume a near correct rotation is known. Then an orthogonal rotation matrix looks like:

$$R = \begin{bmatrix} 1 & -\omega_z & \omega_y \\ \omega_z & 1 & -\omega_x \\ -\omega_y & \omega_x & 1 \end{bmatrix}$$

where ω is the 3D rotation axis and its length is the amount by which to rotate

Using this matrix, iteratively and linearly solve for ω 's and t_{LR} :

$$(R \ \widetilde{x}_L + t_{LR}) - \widetilde{x}_R = 0$$

Limitations:

- 1. ignores normality/scale (fix by re-scaling each iteration)
- 2. assumes good initial guess

How many equations/scene-points are needed?

6 unknowns, 3 equations per scene point, so \geq 2 points

Occlusion Compatible Rendering Order



- Project the new viewpoint onto the original image and divide the image into 1, 2 or 4 "sheets"
- Instantaneous epipole







Correspondence









Epipolar Constraint: reduces correspondence problem to 1D search along *conjugate epipolar lines*



Epipolar Constraint: can be expressed using the *fundamental matrix* F





converging cameras









motion parallel with image plane









Forward motion







Correspondence reduced to looking in a small neighborhood of a line...



Fundamental Matrix



How to compute the fundamental matrix?

- 1. geometric explanation...
- 2. algebraic explanation...





Thus, there is a mapping $x \rightarrow l'$ $\uparrow \qquad \uparrow$ point line





How do you map a point to a line?





Idea:

- We know (x')'s are in a plane
- Define a line by its "perpendicular", then we can use dot product; e.g., $x' \cdot l' = 0$ or $(x' c') \cdot l' = 0$





What is a definition of l' as perpendicular to the pictured epipolar line?

$$l' = (e' - c') \times (x' - c') \quad \longrightarrow \quad l' = e' \times x'$$

(assume all in canonical frame of the right-side camera)



$$l' = e' \times x'$$

Cross product can be expressed using matrix notation:

$$e' \times x' = \begin{bmatrix} 0 & -e'_{z} & e'_{y} \\ e'_{z} & 0 & -e'_{x} \\ -e'_{y} & e'_{x} & 0 \end{bmatrix} \begin{bmatrix} x'_{x} \\ x'_{y} \\ x'_{z} \end{bmatrix}$$

$$e' \times x' = [e']_{\times} x'$$

$$l' = [e']_{\times} x'$$



How do you compute x'?

Use a homography (or projective transformation) to map x to x'

(Homography: maps points in a plane to another plane)

$$x = \begin{bmatrix} x_x \\ x_y \\ 1 \end{bmatrix}, x' = \begin{bmatrix} w'x'_x \\ w'x'_y \\ w' \end{bmatrix}, H = \begin{bmatrix} \ddots & \ddots \\ \vdots & \ddots \\ \vdots & \vdots \end{bmatrix}$$

$$x' = Hx$$







Fundamental Matrix: Algebraic Exp.





Fundamental Matrix: Algebraic Exp.

$$x = PX \qquad X' = ?$$

 $X(t) = P^+x + tc$ where P^+ is the pseudoinverse of P

Why pseudoinverse? Since P not square, pseudoinverse means $PP^+ = I$ but solved as an optimization

Recall $l' = [e']_{\times} x'$

What is x' in terms of x?

(Let's assume t = 0 which means X in on the image plane)

$$x' = \mathcal{P}'P^+x \implies F = [e']_{\times}P'P^+ \implies x'^TFx = 0$$

Epipolar Constraint



Epipolar constraint reduces correspondence problem to 1D search along *conjugate epipolar lines*





Correspondence: Epipolar Geometry

Interesting case: what happens if camera motion is pure translation?



 $P = \begin{bmatrix} I \mid 0 \end{bmatrix} \quad P' = \begin{bmatrix} I \mid t \end{bmatrix}$ $F = [e'] \qquad (H = I)$

If motion parallel to x-axis...

Thus the desire to do image rectification

$$e' = \begin{bmatrix} 1 & 0 & 0 \end{bmatrix}^T$$
$$F = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & -1 \\ 0 & 1 & 0 \end{bmatrix}$$
implies horizontal epipolar line...



Correspondence: Epipolar Geometry



Thus for rectified images, correspondence is reduced to looking in a small neighborhood of a line...

Essential Matrix



 Similar to the fundamental matrix but includes the intrinsic calibration matrix, thus the equation is in terms of the normalized image coordinates, e.g.:

$$x'^T Ex' = 0$$
 and $E = K'^T FK$
essential matrix



Camera geometry known

Correspondence and epipolar geometry known

What is the location of the scene point (scene geometry)?





 $\widetilde{x}_a = M_a \widetilde{X}$ or $\widetilde{x}_b = M_b \widetilde{X}$

Problem? Assumes we know $\widetilde{x} = \begin{bmatrix} x' & y' & w' \end{bmatrix}^T$ But what is the value for w'?

Scene Geometry: Linear Formulation

$$\widetilde{x} = M\widetilde{X} \text{ where } \widetilde{x} = \begin{bmatrix} x' \\ y' \\ w' \end{bmatrix}$$
Recall $\begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x'/w' \\ y'/w' \end{bmatrix}$ where x and y are the observed projections
Let $\widetilde{x} = \begin{bmatrix} sx \\ sy \\ s \end{bmatrix} = \begin{bmatrix} x' \\ y' \\ w' \end{bmatrix}$, thus $s = W'$

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Hence? $sx = m_{11}X + m_{12}Y + m_{13}Z + m_{14}$ $sy = m_{21}X + m_{22}Y + m_{23}Z + m_{24}$

Scene Geometry: Linear Formulation

$$sx = m_{11}X + m_{12}Y + m_{13}Z + m_{14}$$

Given $sy = m_{21}X + m_{22}Y + m_{23}Z + m_{24}$ and N cameras
 $s = m_{31}X + m_{32}Y + m_{33}Z + m_{34}$

For a scene point, how many unknowns? 3+N For a scene point, how many camera 3N≥3+N views needed?

In general, one scene point observed in at least two views is sufficient...





$$\begin{cases} sx = m_{11}X + m_{12}Y + m_{13}Z + m_{14} \\ sy = m_{21}X + m_{22}Y + m_{23}Z + m_{24} \\ s = m_{31}X + m_{32}Y + m_{33}Z + m_{34} \end{cases} x2$$

Scene Geometry: Linear Formulation

$$\begin{bmatrix} m_{11} & m_{12} & m_{13} & -x & 0 \\ m_{21} & m_{22} & m_{23} & -y & 0 \\ m_{31} & m_{32} & m_{33} & -1 & 0 \\ m'_{11} & m'_{12} & m'_{13} & 0 & -x' \\ m'_{21} & m'_{22} & m'_{23} & 0 & -y' \\ m'_{31} & m'_{32} & m'_{33} & 0 & -1 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ s \\ s' \end{bmatrix} = \begin{bmatrix} -m_{14} \\ -m_{24} \\ -m'_{14} \\ -m'_{24} \\ -m'_{34} \end{bmatrix}$$

Cameras *M* and *M*'

$$\begin{cases} sx = m_{11}X + m_{12}Y + m_{13}Z + m_{14} \\ sy = m_{21}X + m_{22}Y + m_{23}Z + m_{24} \\ s = m_{31}X + m_{32}Y + m_{33}Z + m_{34} \end{cases} x2$$

Scene Geometry: Nonlinear Form.

- Remember "Bundle Adjustment"
 - Given initial guesses, use nonlinear least squares to refine/compute the calibration parameters
 - Simple but good convergence depends on accuracy of initial guess

Scene Geometry: Nonlinear Form.

Recall

$$E = \frac{1}{mn} \sum_{ij} \left[(x_{ij} - \frac{m_{i1} \cdot \widetilde{X}_j}{m_{i3} \cdot \widetilde{X}_j})^2 + (y_{ij} - \frac{m_{i2} \cdot \widetilde{X}_j}{m_{i3} \cdot \widetilde{X}_j})^2 \right]$$

Goal is $E \to 0$

For scene geometry, \widetilde{X} are the unknowns...

Example Result



• Using dense feature-based stereo

