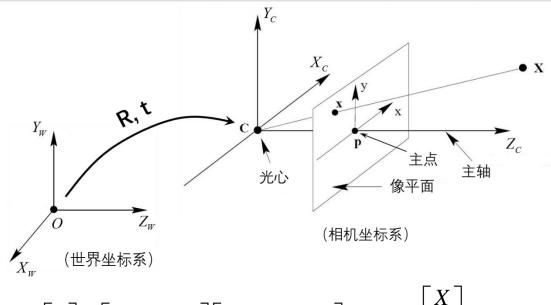
三维视觉初步—— 稀疏重建与相机姿态估计

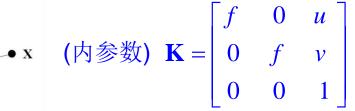
陶文兵

华中科技大学人工智能与自动化学院 2022-08-07

相机模型(Camera Model)

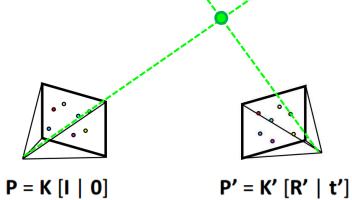






$$($$
外参数 $)$ \mathbf{R} \mathbf{R} \mathbf{t}

$$\begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} f & 0 & u \\ 0 & f & v \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} R & t \\ 0^T & 1 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$
$$\mathbf{x} = \mathbf{K} \begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix} \mathbf{X} = \mathbf{P} \mathbf{X}$$

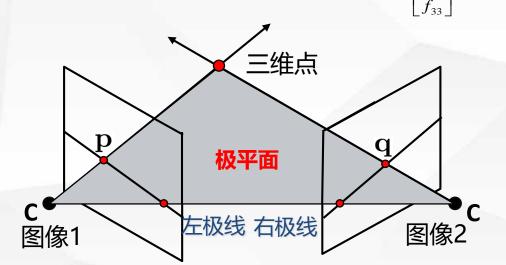


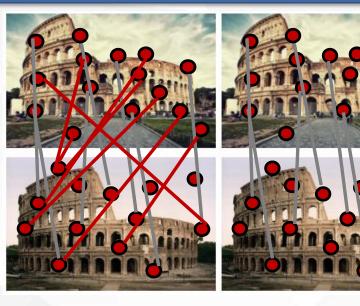
基础矩阵 (Fundamental Matrix)



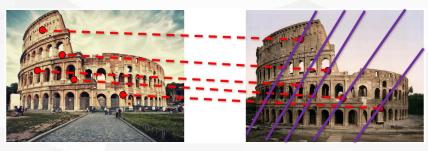


$$\begin{bmatrix} u_{1}u_{1} & v_{1}u_{1} & u_{1} & u_{1}v_{1} & v_{1}v_{1} & v_{1} & u_{1} & v_{1} & 1 \\ u_{2}u_{2} & v_{2}u_{2} & u_{2} & u_{2}v_{2} & v_{2}v_{2} & v_{2} & u_{2} & v_{2} & 1 \\ \vdots & \vdots \\ u_{n}u_{n} & v_{n}u_{n} & u_{n} & u_{n}v_{n} & v_{n}v_{n} & v_{n} & u_{n} & v_{n} & 1 \end{bmatrix} \begin{bmatrix} f_{12} \\ f_{22} \\ f_{22} \\ f_{33} \\ f_{34} \end{bmatrix}$$





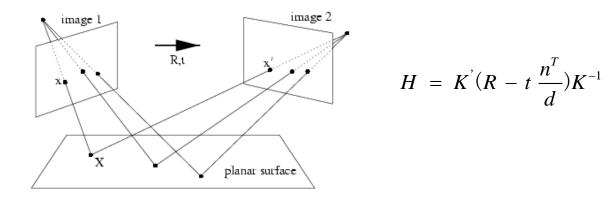
=0



单应矩阵(Homography Matrix)



$$\begin{bmatrix} x' \\ y' \\ z' \end{bmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \boldsymbol{H} \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

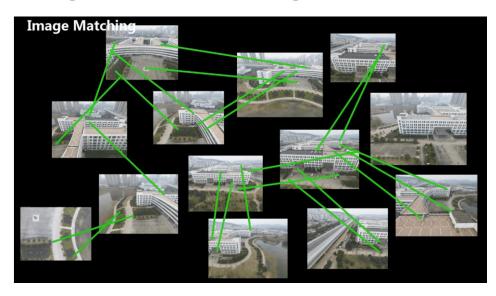


If n 3D points lie on the same planar surface, then the n matchs corresponding these 3D points satisfiy one homography transformation.

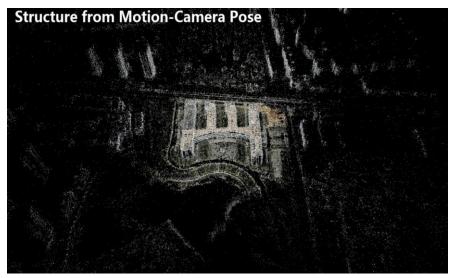
Sparse Reconstruction and Pose Estimation



Image Feature Matching



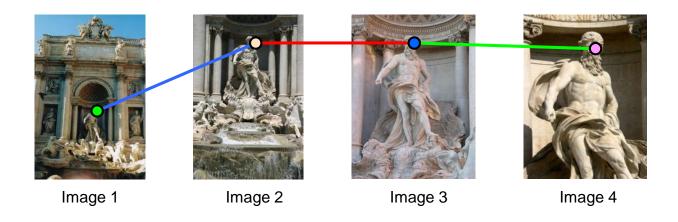
Camera Pose Estimation



Pose estimation—Incremental Structure from Motion (SFM)



Build tracks from matches



- Link up matches between pairs of images into tracks between multiple images
- Each track corresponds to a 3D point

Pose estimation—— Incremental Structure from Motion(SFM)

THE PARTY OF THE P

- Initialization
 - 1. Choose two views



lots of correspondences



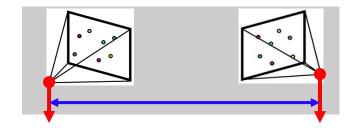
wide baseline

narrow baseline



few correspondences

- They have the most number of correspondences
- They have wide baselines









lots of correspondences



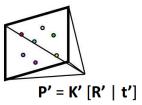
- Initialization
 - 1. Choose two views
 - 2. Estimate relative pose using two-view geometry



- Camera intrinsics knownEssential matrix, E (5 points)
- Camera intrinsics unknown Fundamental matrix, F (7 points)

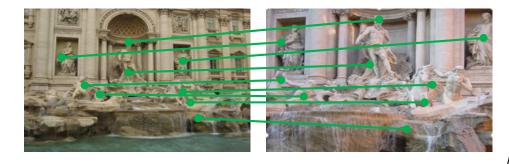


$$P = K [I \mid 0]$$

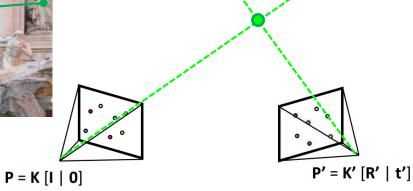




- Initialization
 - 1. Choose two views
 - 2. Estimate relative pose using two-view geometry
 - 3. Triangulate inlier correspondences

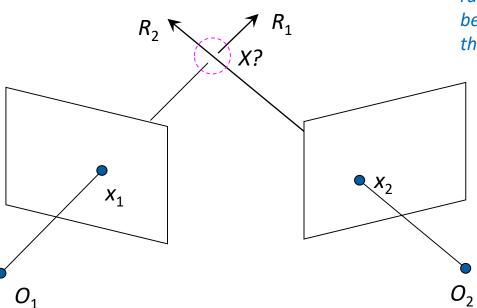


Given projections of a 3D point in two or more images (with known camera matrices), find the coordinates of the point





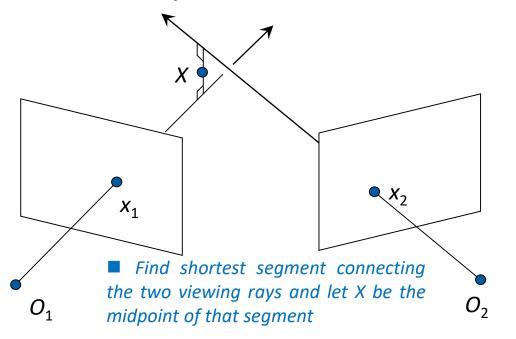
- Initialization
 - 1. Choose two views
 - 2. Estimate relative pose using two-view geometry
 - 3. Triangulate inlier correspondences



We want to intersect the two visual rays corresponding to x1 and x2, but because of noise and numerical errors, they don't meet exactly

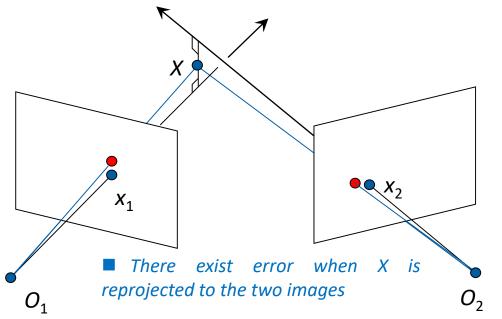


- Initialization
 - 1. Choose two views
 - 2. Estimate relative pose using two-view geometry
 - 3. Triangulate inlier correspondences
 - 4. Bundle adjustment



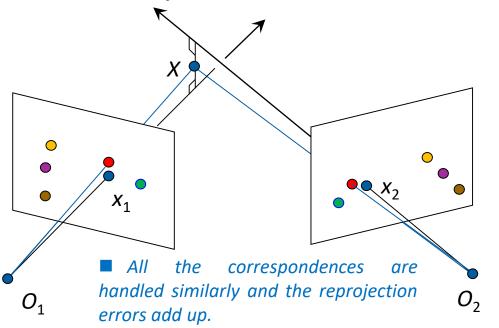


- Initialization
 - 1. Choose two views
 - 2. Estimate relative pose using two-view geometry
 - 3. Triangulate inlier correspondences
 - 4. Bundle adjustment



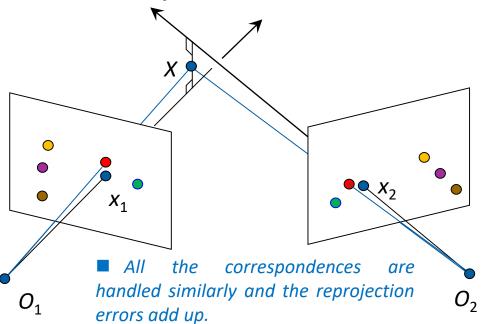


- Initialization
 - 1. Choose two views
 - 2. Estimate relative pose using two-view geometry
 - 3. Triangulate inlier correspondences
 - 4. Bundle adjustment





- Initialization
 - 1. Choose two views
 - 2. Estimate relative pose using two-view geometry
 - 3. Triangulate inlier correspondences
 - 4. Bundle adjustment



- refine 3D points
- refine camera parameters
- Minimize reprojection error:

$$E(\mathbf{P}, \mathbf{X}) = \sum_{i=1}^{m} \sum_{j=1}^{n} w_{ij} D(\mathbf{x}_{ij}, \mathbf{P}_{i} \mathbf{X}_{j})^{2}$$

$$\mathbf{W}_{ij} \quad \text{indicator variable for visibility}$$
of point \mathbf{X}_{j} in camera \mathbf{P}_{i}

- Non-linear least squares problem
- Minimizing by Levenberg-Marquardt (LM), is called Bundle Adjustment (BA)