# 3D Computer Vision Syllabus

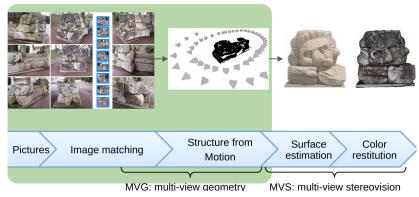
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### Structure from Motion

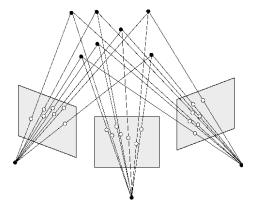


We will cover the first part of this Stereo pipeline, MVG, and first part of MVS but only with two views.

- Output of MVG: point cloud and camera positions and orientations
- Output of MVS: a mesh surface with colors

## Bundle Adjustment

- Observed points in images are central projections of unknown 3D points.
- ► Each image provides a bundle of half-lines containing the 3D points.
- ▶ We must orient the bundles so that the lines intersect.



Source: Manolis Lourakis

# One equation to rule them all

$$\arg \min_{\{R_i\}, \{T_i\}, \{\mathsf{X}_j\}} \sum_{i,j} \epsilon_{ij} \ d\left(\mathsf{x}_{ij}, \mathsf{\Pi}_i \big(R_i \mathsf{X}_j + T_i\big)\right)^2.$$

- $\triangleright$   $x_{ij}$ : point number j (among n) in view i (among m), pixel coordinates.
- ▶  $\epsilon_{ij} \in \{0,1\}$ : visibility of point j in view i.
- d: Euclidean distance in image.
- $ightharpoonup \Pi_i$ : 2D projection operator in view *i* from 3D.

#### Unknowns:

- $\triangleright$  X<sub>j</sub>: 3D point in a fixed coordinate frame.
- R<sub>i</sub>: rotation matrix of coordinate frame wrt view i.
- $ightharpoonup T_i$ : position of origin O of coordinate frame wrt view i.

## Session 1

$$\arg \min_{\{R_i\}, \{T_i\}, \{\mathsf{X}_j\}} \sum_{i,j} \epsilon_{ij} \ d\left(\mathsf{x}_{ij}, \mathsf{\Pi}_i \big(R_i \mathsf{X}_j + T_i\big)\right)^2.$$

- Expression of projection model Π<sub>i</sub> (i.e., camera model)
- ▶ Case m = 2,  $T_L = T_R = 0$  or all  $X_i$  on a single plane:

$$\forall i \in \{L, R\}, \forall j, \ d\left(\mathsf{x}_{ij}, \mathsf{\Pi}_i(R_i \mathsf{X}_j + 0)\right) = 0.$$

Then  $x_{Rj} = H_{\Pi_R,R_R}(x_{Lj})$ , and the homography H can be determined from four pairs  $\{x_{Li}, x_{Ri}\}$ . Application: panorama.

► Case m = 1,  $n \ge 6$  pairs  $(x_j, X_j)$ , recover all info about the camera (camera calibration from 3D rig):

$$\arg\min_{\Pi,R,T}\sum_{i}d\left(\mathbf{x}_{ij},\Pi(\mathbf{RX}_{j}+\mathbf{T})\right)^{2}.$$

► Case  $m \ge 3$ , single camera  $\Pi$ , 4 (or more) pairs  $(x_j, X_j)$ , all  $X_j$  on a common plane (camera calibration from 2D rig):

$$\arg\min_{\Pi,\{R_i\},\{T_i\}}\sum_{i:}d\left(\mathbf{x}_{ij},\frac{\Pi(R_i\mathbf{X}_j+T_i)}{}\right)^2.$$

$$\arg\min_{\{R_i\},\{T_i\},\{\mathsf{X}_j\}}\sum_{i,j}\epsilon_{ij}\;d\left(\mathsf{x}_{ij},\mathsf{\Pi}_i(R_i\mathsf{X}_j+T_i)\right)^2.$$

Case m = 2 views.

► From  $n \ge 5$  pairs  $\{x_{Lj}, x_{Rj}\}$  and known  $\Pi_L$ ,  $\Pi_R$ , deduce a bilinear constraint  $\forall j, E_{R_R, T_R}(x_{Lj}, x_{Rj}) = 0$  (essential matrix). Recover  $(R_R, T_R)$  from E.

$$\arg\min_{\{R_i\},\{T_i\},\{X_j\}}\sum_{i\in\{L,R\},i}\epsilon_{ij}\;d\left(\mathsf{x}_{ij},\mathsf{\Pi}_i(R_i\mathsf{X}_j+T_i)\right)^2.$$

- ► From  $n \ge 7$  pairs  $\{x_{Lj}, x_{Rj}\}$  and unknown  $\Pi_L$ ,  $\Pi_R$ , deduce a bilinear constraint  $\forall j, F(x_{Li}, x_{Ri}) = 0$  (fundamental matrix).
- From *n* pairs  $\{x_{Lj}, x_{Rj}\}$ , maximize the number of matching pairs up to tolerance  $\tau \geq 0$  (RANSAC algorithm)

$$\max \sum_{i \in \{L,R\},j} \epsilon_{ij}$$
s.t.  $\exists \Pi_L, \Pi_R, \{R_i\}, \{T_i\}, \{X_i\}, d\left(\mathbf{x}_{ii}, \Pi_i(R_i \mathbf{X}_i + T_i)\right) \leq \tau$ .

## Sessions 3&4

$$\arg\min_{\{R_i\},\{T_i\},\{\mathsf{X}_j\}}\sum_{i,i}\epsilon_{ij}\;d\left(\mathsf{x}_{ij},\mathsf{\Pi}_i(R_i\mathsf{X}_j+T_i)\right)^2.$$

Case m=2 views,  $R_L=R_R=I$ ,  $T_L=0$ ,  $T_R=Be_1$  (rectified pair).

- ▶ Go from generic poses to rectified pair with fundamental matrix F.
- ▶ For all  $x_{Lj}$ ,  $j \in \{\text{pixels of left image}\}$ , find corresponding pixel  $x_{Rj}$ , hence  $X_j$  by triangulation.
- ► Two categories of methods: local/global.

## Session 5

$$\arg\min_{\{R_i\},\{T_i\},\{X_j\}}\sum_{i,j}\epsilon_{ij}\;d\left(\mathsf{x}_{ij},\mathsf{\Pi}_i(R_i\mathsf{X}_j+T_i)\right)^2.$$

Case  $m \geq 3$ , multi-view stereovision.

- ► For m = 3, trilinear constraints  $\mathcal{T}(x_{1j}, x_{2j}, x_{3j}) = 0$ . Recovery of tensor  $\mathcal{T}$ .
- Recover  $R_{m+1}$  and  $T_{m+1}$  from known pairs  $(x_{m+1j}, X_j)$  (resection, PnP Perspective from n Points):

$$\arg\min_{R_{m+1},T_{m+1}}\sum_{j}\epsilon_{m+1j}\;d\left(\mathbf{x}_{ij},\Pi_{m+1}(R_{m+1}\mathbf{X}_{j}+T_{m+1})\right)^{2}.$$

- ⇒ incremental multi-view pipeline.
- ▶ Special case  $T_i = B_i v$  with v a fixed vector and known  $B_i$ :

$$\arg\min_{\{R_i\},\{X_j\}}\sum_{i,j}\epsilon_{ij}\ d\left(\mathbf{x}_{ij},\Pi_i(R_i\mathbf{X}_j+B_i\mathbf{v})\right)^2.$$

⇒ Light field imagery, epipolar plane imagery

## Session 6

$$\arg\min_{\{R_i\},\{T_i\},\{\mathsf{X}_j\}}\sum_{i,i}\epsilon_{ij}\;d\left(\mathsf{x}_{ij},\mathsf{\Pi}_i(R_i\mathsf{X}_j+T_i)\right)^2.$$

Case m=2 views, find interest points  $\{p_{Lk}\}$  and  $\{p_{Rk'}\}$ . Deduce pairs j=(k,k') of matching points  $(x_{Lj},x_{Rj})$  with  $x_{Lj}=p_{Lk}$  and  $x_{Rj}=p_{Rk'}$ .