

Computer Vision: Lecture 10

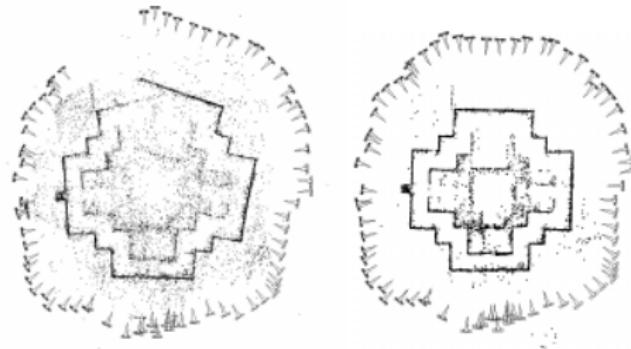
2023-11-30

Today's Lecture

Non-Sequential Methods for SfM

- Rotation averaging and translation registration
- Project presentation

Why non-sequential SfM: drift



Minimizing the Reprojection Error

Main goal

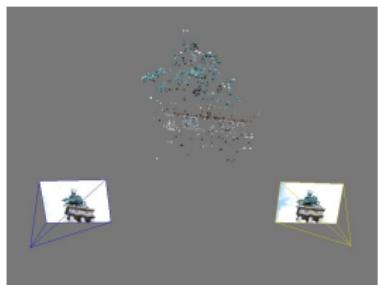
For given $\{\mathbf{x}_{ij}\}$ and $\{m_{ij}\}$ find a minimizer

$$\sum_{i,j} m_{ij} \|\mathbf{x}_{ij} - \pi(\mathbf{P}_i \mathbf{X}_j)\|^2 \rightarrow \min_{\{\mathbf{P}_i\}, \{\mathbf{X}_j\}}$$

Local optimization needs good starting point

$N = 2$: Recall lecture 5+6

- Compute F/E
- Compute a pair of camera matrices: $\mathbf{P}_1 = (\mathbf{I} \mid \mathbf{0})$, $\mathbf{P}_2 = ([\mathbf{e}_2]_\times \mathbf{F} \mid \mathbf{e}_2)$
- Triangulate \mathbf{X}_j using DLT



Minimizing the Reprojection Error

Main goal

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$$\sum_{i,j} m_{ij} \|\mathbf{x}_{ij} - \pi(\mathbf{P}_i \mathbf{X}_j)\|^2 \rightarrow \min_{\{\mathbf{P}_i\}, \{\mathbf{X}_j\}}$$

Local optimization needs good starting point

$N > 2$: Sequential / incremental SfM



How do we compute the entire reconstruction?

- ① For a pair of images compute initial reconstruction ($N = 2$)
- ② For a new image viewing some of the previously reconstructed scene points, find the camera matrix (via DLT)
- ③ Compute new scene points using triangulation (DLT)
- ④ If there are more cameras/images goto step 2

Minimizing the Reprojection Error

Main goal

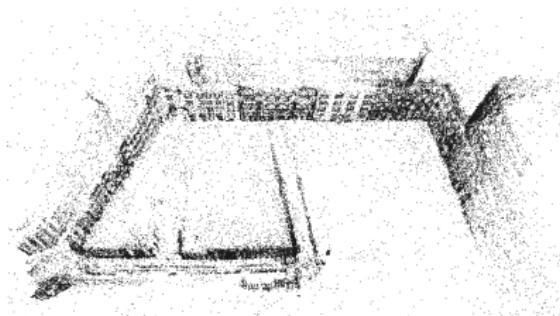
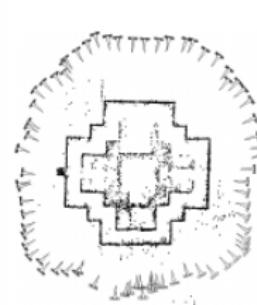
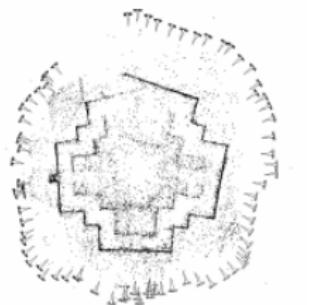
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Local optimization needs good starting point

$N > 2$: Sequential / incremental SfM

Main drawback: drift due to error accumulation



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Main goal

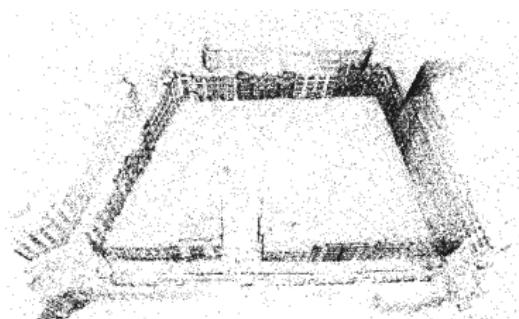
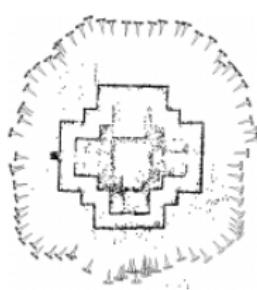
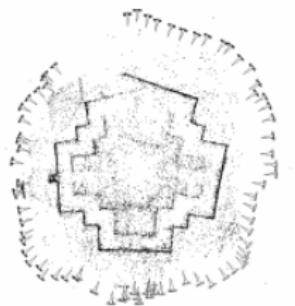
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Minimizing the Reprojection Error

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For given $\{\mathbf{x}_{ij}\}$ and $\{m_{ij}\}$ find a minimizer

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Local optimization needs good starting point

Why is this objective hard to minimize?

- Bilinear, non-convex terms $\mathbf{P}_i \mathbf{X}_j$
- Perspective division $\pi(\mathbf{X}) = \frac{1}{X_3} \begin{pmatrix} X_1 \\ X_2 \end{pmatrix}$
- Calibrated SfM: constraints $\mathbf{R}_i \in SO(3)$

Minimizing the Reprojection Error

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Minimizing the Reprojection Error

Question

We have a good idea how to compute F/E robustly for image pairs:

RANSAC + 5/7/8-point method

Can we use this pairwise information?

Issues:

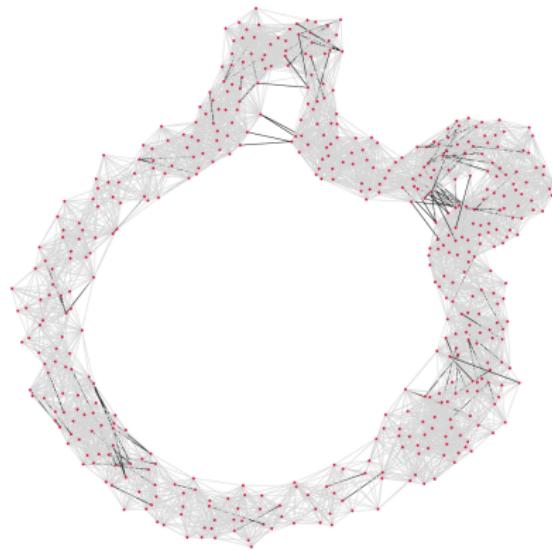
- Ambiguity: projective ambiguity (F), similarity transformation (E)
- Perceptual aliasing

Calibrated Structure and Motion

- Estimate E_{ij} between images i and j if possible



≈ 500 images



Camera graph

Calibrated Structure and Motion

- Each E_{ij} yields relative rotations R_{ij} and translations T_{ij}
- Pair of camera matrices in local coordinate system

$$\tilde{P}_{ij,0} = (\mathbb{I} \mid \mathbf{0}) \quad \tilde{P}_{ij,1} = (R_{ij} \mid T_{ij})$$

- Camera matrices in common coordinate system $P_i = (R_i \mid T_i)$
- Transformation from P_i/P_j to a canonical pair $\tilde{P}_{ij,0}/\tilde{P}_{ij,1}$

$$\begin{aligned}(R_i \mid T_i) \begin{pmatrix} R_i^\top & -R_i^\top T_i \\ \mathbf{0}^\top & 1 \end{pmatrix} &= (\mathbb{I} \mid \mathbf{0}) \\ (R_j \mid T_j) \begin{pmatrix} R_i^\top & -R_i^\top T_i \\ \mathbf{0}^\top & 1 \end{pmatrix} &= (R_j R_i^\top \mid T_j - R_j R_i^\top T_i) \\ &\stackrel{!}{=} (R_{ij} \mid T_{ij})\end{aligned}$$

- Therefore $R_{ij} = R_j R_i^\top$ and $T_{ij} = T_j - R_j R_i^\top T_i$

Calibrated Structure and Motion

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- Therefore $R_{ij} = R_j R_i^\top$ and $T_{ij} = T_j - R_j R_i^\top T_i$

Calibrated Structure and Motion

Rotation averaging

For given relative rotation matrices $\{R_{ij}\}$ determine $\{R_i\}$ such that

$$R_{ij} \approx R_j R_i^\top \quad \text{or} \quad R_{ij} R_i - R_j \approx 0$$

- Method 0: set $R_1 = I$ and chain rotations

$$R_j \leftarrow R_{ij} R_i \quad \text{if } R_j \text{ is not assigned yet}$$

- Method 1: null-space method

Place R_{ij} and $-I$ in the columns corresponding to R_i and R_j

$$\underbrace{\begin{pmatrix} R_{ij} & \vdots & -I \\ \vdots & \ddots & \vdots \\ R_{ij} & \vdots & -I \end{pmatrix}}_{=A} \begin{pmatrix} R_1 \\ \vdots \\ R_2 \end{pmatrix} = 0_{3N \times 3}$$

Extract 3 rightmost columns of V , $U\Sigma V^\top = \text{SVD}(A)$

- Method 2: solve a semi-definite program (SDP)

Calibrated Structure and Motion

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Calibrated Structure and Motion

Translation registration I

For fixed rotation matrices $\{R_i\}$ and given relative translations $\{T_{ij}\}$ determine translations $\{T_i\}$ such that

$$T_{ij} \sim T_j - R_j R_i^\top T_i$$

Problem 1: T_{ij} are only known up to scale.

$$T_{ij} \times (T_j - R_j R_i^\top T_i) \approx \mathbf{0}$$

Problem 2: numerical stability when $R_j R_i^\top \approx I$ (e.g. linear motion)

Translation registration II

For fixed rotation matrices $\{R_i\}$ determine translations $\{T_i\}$ and scene points $\{X_j\}$ such that

$$x_{ij} \approx \pi(R_i X_j + T_i)$$

Calibrated Structure and Motion

Translation registration II

For fixed rotation matrices $\{\mathbf{R}_i\}$ determine translations $\{\mathbf{T}_i\}$ and scene points $\{\mathbf{X}_j\}$ such that

$$\mathbf{x}_{ij} \approx \pi(\mathbf{R}_i \mathbf{X}_j + \mathbf{T}_i)$$

- Method 1: linear DLT solution

$$\mathbf{x}_{ij} \times (\mathbf{R}_i \mathbf{X}_j + \mathbf{T}_i) = \mathbf{0} \iff [\mathbf{x}_{ij}]_{\times} \mathbf{R}_i \mathbf{X}_j + [\mathbf{x}_{ij}]_{\times} \mathbf{T}_i$$

- Homogeneous least-squares instance

$$\begin{pmatrix} & & & \\ & \vdots & & \\ [\mathbf{x}_{ij}]_{\times} & & [\mathbf{x}_{ij}]_{\times} \mathbf{R}_i & \\ & \vdots & & \\ & & & \end{pmatrix} \begin{pmatrix} \vdots \\ \mathbf{T}_i \\ \vdots \\ \mathbf{X}_j \\ \vdots \end{pmatrix} = \mathbf{0}$$

Calibrated Structure and Motion

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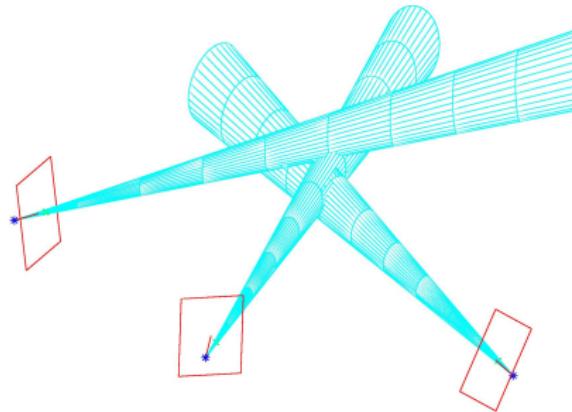
Calibrated Structure and Motion

Translation registration II

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Method 2:



Translation registration II

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Method 2:

- Constraint $\|\mathbf{x}_{ij} - \pi(\mathbf{R}_i \mathbf{X}_j + \mathbf{T}_i)\| \leq b$ is essentially convex in \mathbf{X}_j and \mathbf{T}_i

$$\|\mathbf{x}_i - \pi(\mathbf{R}_i \mathbf{X}_j + \mathbf{T}_i)\| \leq b \wedge (\mathbf{R}_i \mathbf{X}_j + \mathbf{T}_i)_3 \geq 0$$

$$\iff \|(\mathbf{R}_i \mathbf{X}_j + \mathbf{T}_i)_3 \mathbf{x}_i - (\mathbf{R}_i \mathbf{X}_j + \mathbf{T}_i)_{1,2}\| - (\mathbf{R}_i \mathbf{X}_j + \mathbf{T}_i)_3 b \leq 0 \wedge (\mathbf{R}_i \mathbf{X}_j + \mathbf{T}_i)_3 \geq 0$$

Convex cone

- Linear or SOCP constraint (recall lecture 9)
- Convex feasibility problem
- Bisection search (lecture 9) to determine $b \geq 0$

Calibrated Structure and Motion

Translation registration II

For fixed rotation matrices $\{R_i\}$ determine translations $\{T_i\}$ and scene points $\{X_j\}$ such that

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Method 3:

- Shorthand notation: $X'_{ij} = R_i X_j + T_i$ linear in X_j and T_i
- Take a closer look at the squared reprojection error

$$\|x_{ij} - \pi(X'_{ij})\|^2 = \frac{\left\| Z'_{ij} x_{ij} - \begin{pmatrix} X'_{ij} \\ Y'_{ij} \end{pmatrix} \right\|^2}{(Z'_{ij})^2} \quad X'_{ij} = (X'_{ij}, Y'_{ij}, Z'_{ij})^\top$$

- Replace with convex approximation

$$\frac{\left\| Z'_{ij} x_{ij} - \begin{pmatrix} X'_{ij} \\ Y'_{ij} \end{pmatrix} \right\|^2}{Z'_{ij}} \quad \text{s.t. } Z'_{ij} \geq 0$$

- Convex optimization problem

Calibrated Structure and Motion

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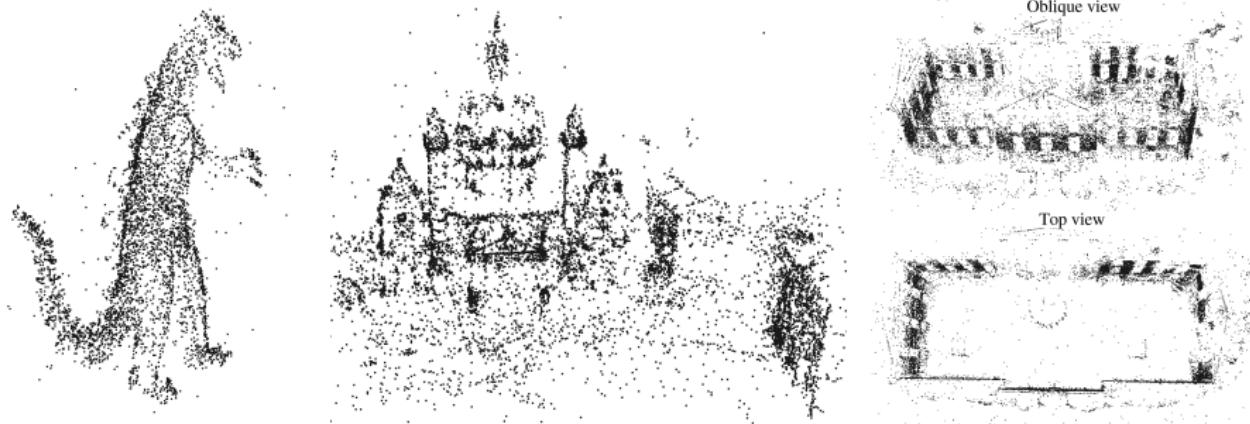
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Calibrated Structure and Motion

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Rotation averaging (method 1) + translation registration (method 3)

Calibrated Structure and Motion

Discussion

Non-sequential SfM (rotation averaging and translation registration) have shortcomings.

- Often not very scalable
 - Large non-linear convex programs to solve
 - Can only handle a very limited amount of outliers
- For city-scale 3D models
 - Combine with “bottom-up” approach: merge smaller sub-models
 - Exploit redundancy in images: apply non-sequential SfM on sparser image set
- Provides not very accurate initial models
 - Average reprojection error ≈ 5 pixels
 - Can be a problem for (robust) bundle adjustment

Upcoming

Project presentation