

Computer Vision: Lecture 4

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Today's Lecture

Keypoint detection and Matching.

- Repetition: DLT
- Triangulation
- Homography Estimation
- Panoramas



Repetition: Direct Linear Transform - DLT

Algorithm for solving

$$\min_{\|v\|^2=1} \|Mv\|^2.$$

- 1 Compute the factorization

$$M = USV^T$$

(in Matlab).

- 2 Select the solution

$$v = \text{last column of } V.$$

Can solve homogeneous least squares problems:

Ex. The resection problem: Find P and λ_i

$$\lambda_i \mathbf{x}_i \approx P \mathbf{X}_i \quad \text{for all } i.$$



Triangulation

Known

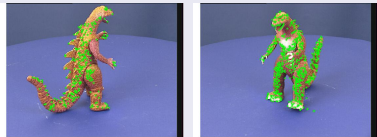
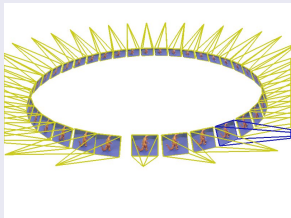
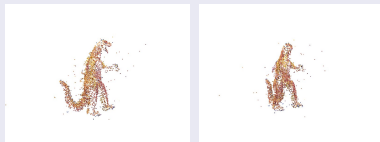


Image points $\{\mathbf{x}_{ij}\}$.



Camera matrices P_j

Sought



3D points $\mathbf{X}_i$, such that

$$\lambda_{ij}\mathbf{x}_{ij} = P_j\mathbf{X}_i$$



Triangulation

Fixed cameras: Determine one 3D point at a time.

Problem Formulation

Given measured projections $\mathbf{x}_i$ and known camera matrices P_i , $i = 1, \dots, n$ compute the corresponding scene point $\mathbf{X}$. Solve

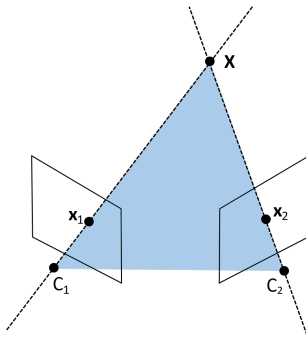
$$\lambda_i \mathbf{x}_i = P_i \mathbf{X} \quad i = 1, \dots, n$$

$3n$ equations, $3 + n$ unknowns. Need $3n \geq 3 + n \Rightarrow n \geq 2$ points.



Triangulation Geometric Interpretation

Two cameras:

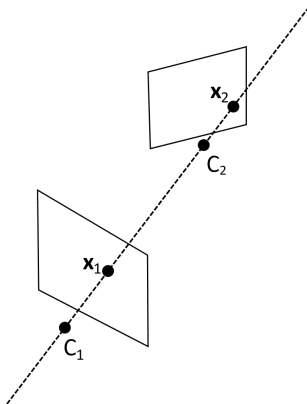


The 3D point is the intersection of the viewing rays.



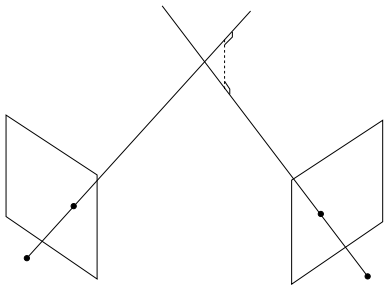
Degenerate Configurations

If all camera centers and the unknown 3D point X are on a line, X cannot be uniquely determined.



Estimation with Noise using DLT

Viewing rays may not intersect in 3D.



DLT

Find the least squares solution of

$$\lambda_1 \mathbf{x}_1 = P_1 \mathbf{X}$$

$$\lambda_2 \mathbf{x}_2 = P_2 \mathbf{X}$$

$$\vdots$$

In matrix form:

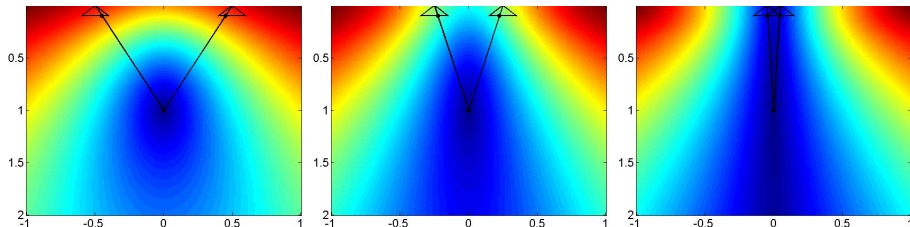
$$\underbrace{\begin{bmatrix} P_1 & -\mathbf{x}_1 & 0 & \dots \\ P_2 & 0 & -\mathbf{x}_2 & \dots \\ \vdots & \vdots & \vdots & \vdots \end{bmatrix}}_{:=M} \begin{bmatrix} \mathbf{X} \\ \lambda_1 \\ \lambda_2 \\ \vdots \end{bmatrix} = 0.$$

Near Degenerate Configurations

$$\min_{\|v\|^2=1} \|Mv\|^2 = \min_X f(X)$$

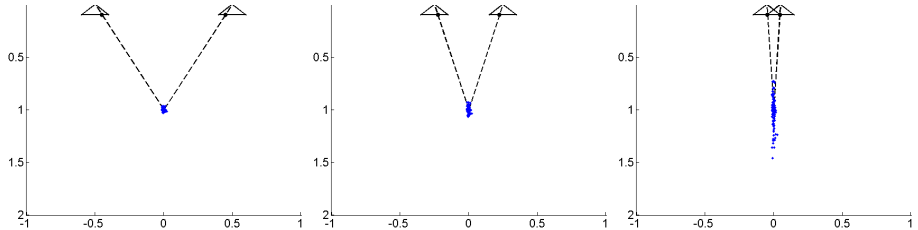
Reduced DLT objective (with known X):

$$f(X) = \min_{\lambda_1^2 + \lambda_2^2 + \|X\|^2 \gamma^2 = 1} \left\| M \begin{bmatrix} \gamma X \\ \lambda_1 \\ \lambda_2 \end{bmatrix} \right\|^2.$$



Near Degenerate Configurations

DLT estimations with noise



Homography Estimation

Problem Formulation

Given 2D points $\mathbf{x}_i$ and corresponding points $\mathbf{y}_i$ related by a projective transformation find H such that

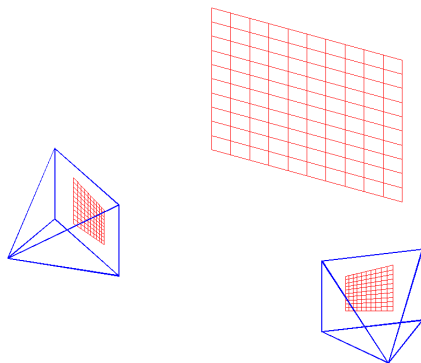
$$\lambda_i \mathbf{y}_i = H \mathbf{x}_i, \quad i = 1, \dots, N.$$

$3N$ equations, $8 + N$ unknowns

Need $3N \geq 8 + N \Rightarrow N \geq 4$ point correspondences.



Homography Estimation Examples



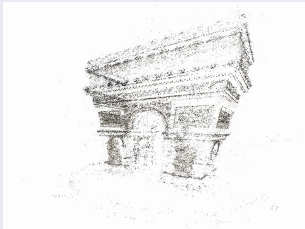
Two images of a plane are related by a $\mathbb{R}^2 \rightarrow \mathbb{R}^2$ homography.



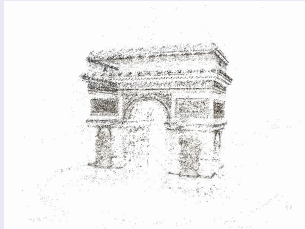
Homography Estimation Examples

Uncalibrated solutions to structure from motion are related by $\mathbb{P}^3 \rightarrow \mathbb{P}^3$ homographies.

Projective



Euclidean



Homography Estimation with Exact Measurements

In general a set of points in $\mathbb{P}^n$ are called **projectively independent** if they have homogeneous coordinates that are linearly independent as vectors in $\mathbb{R}^{n+1}$.

A set of $n + 2$ points in $\mathbb{P}^n$ is called a **projective basis** if no subset of $n + 1$ points is projectively dependent.

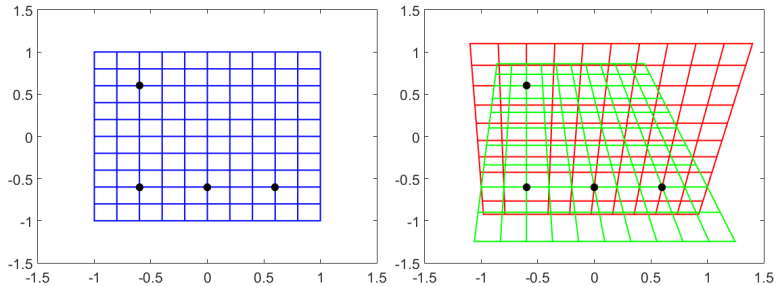
A projective transformation $\mathbb{P}^n \rightarrow \mathbb{P}^n$ is uniquely determined by the mapping of the $n + 2$ points of a projective basis.

Ex1. $\mathbb{P}^2 \rightarrow \mathbb{P}^2$: 4 points, no 3 on a line.

Ex2. $\mathbb{P}^3 \rightarrow \mathbb{P}^3$: 5 points, no 4 on a plane.



Degenerate Cases



Homography Estimation with Noise

DLT

If

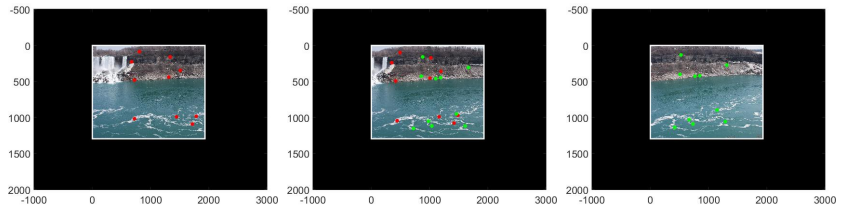
$$H = \begin{bmatrix} H_1^T \\ H_2^T \\ H_3^T \end{bmatrix} \quad \text{and} \quad \mathbf{y}_i = \begin{bmatrix} x_i \\ y_i \\ 1 \end{bmatrix}$$

solve

$$\begin{bmatrix} \mathbf{x}_1^T & 0 & 0 & -x_1 & 0 & \dots \\ 0 & \mathbf{x}_1^T & 0 & -y_1 & 0 & \dots \\ 0 & 0 & \mathbf{x}_1^T & -1 & 0 & \dots \\ \mathbf{x}_2^T & 0 & 0 & 0 & -x_2 & \dots \\ 0 & \mathbf{x}_2^T & 0 & 0 & -y_2 & \dots \\ 0 & 0 & \mathbf{x}_2^T & 0 & -1 & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \end{bmatrix} \begin{bmatrix} H^1 \\ H^2 \\ H^3 \\ \lambda_1 \\ \lambda_2 \\ \vdots \end{bmatrix} = 0.$$



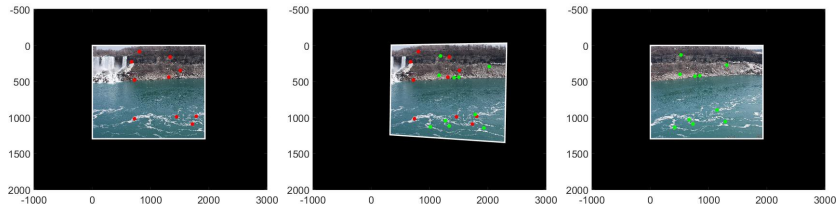
Panoramas and Compositions



H_{21} estimated from green matches, H_{32} estimated from red matches.



Panoramas and Compositions

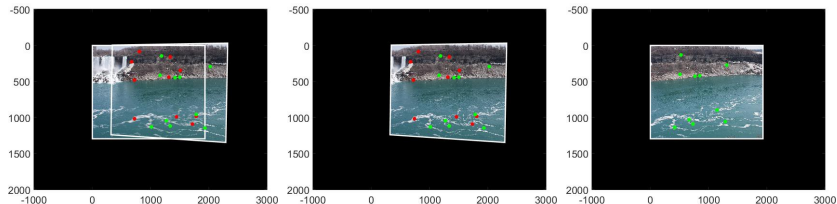


H_{21} estimated from green matches, H_{32} estimated from red matches.

- Transform image 2 using H_{21} .



Panoramas and Compositions

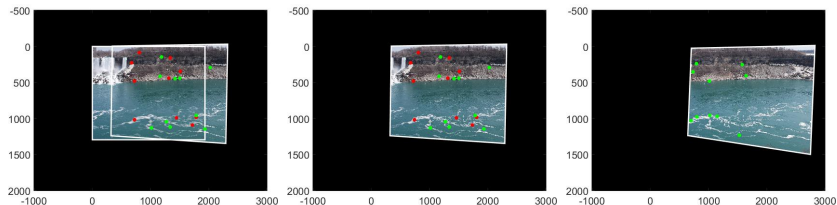


H_{21} estimated from green matches, H_{32} estimated from red matches.

- Transform image 2 using H_{21} .
- Stitch.



Panoramas and Compositions

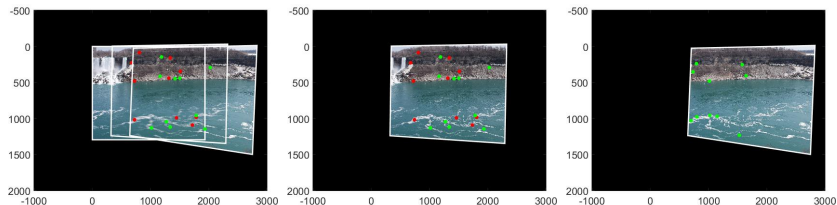


H_{21} estimated from green matches, H_{32} estimated from red matches.

- Transform image 2 using H_{21} .
- Stitch.
- Transform image 2 using $H_{31} = H_{21}H_{32}$.



Panoramas and Compositions



H_{21} estimated from green matches, H_{32} estimated from red matches.

- Transform image 2 using H_{21} .
- Stitch.
- Transform image 2 using $H_{31} = H_{21}H_{32}$.
- Stitch.

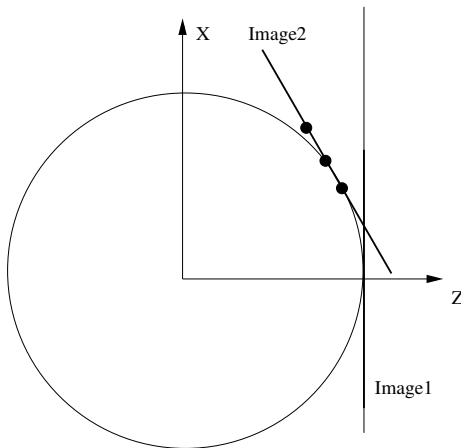


Panoramas and Compositions



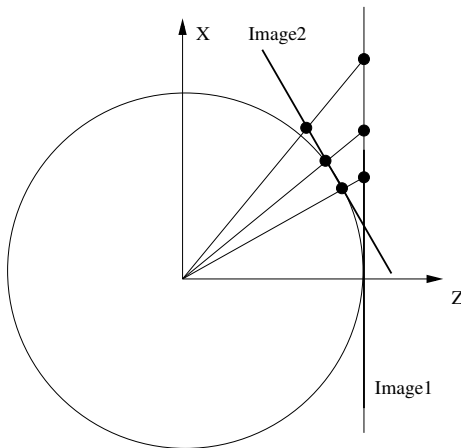
Panoramas

For calibrated cameras:



Panoramas

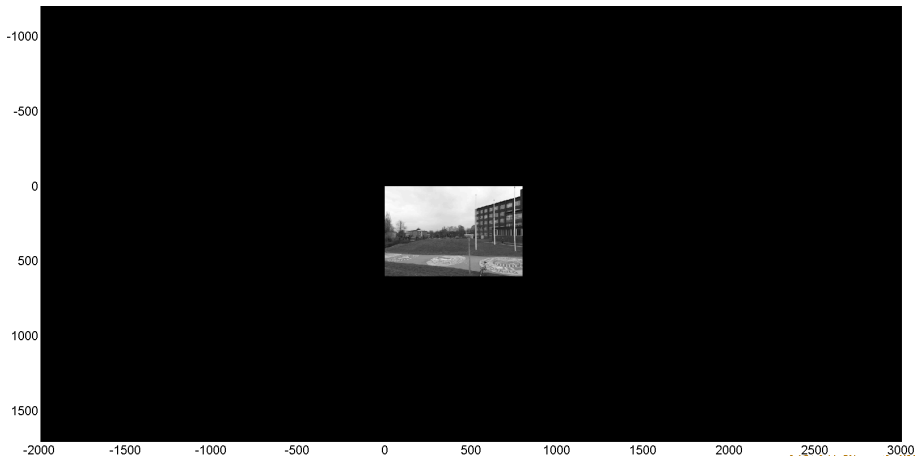
For calibrated cameras:



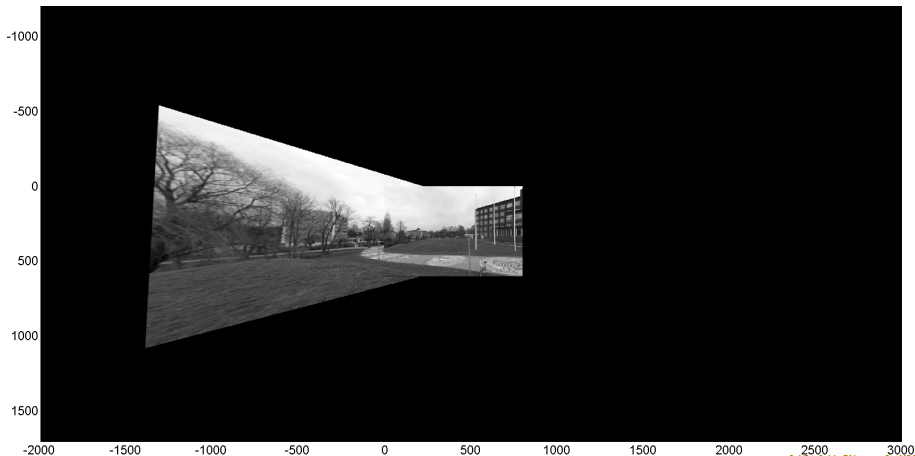
Panoramas



Panoramas



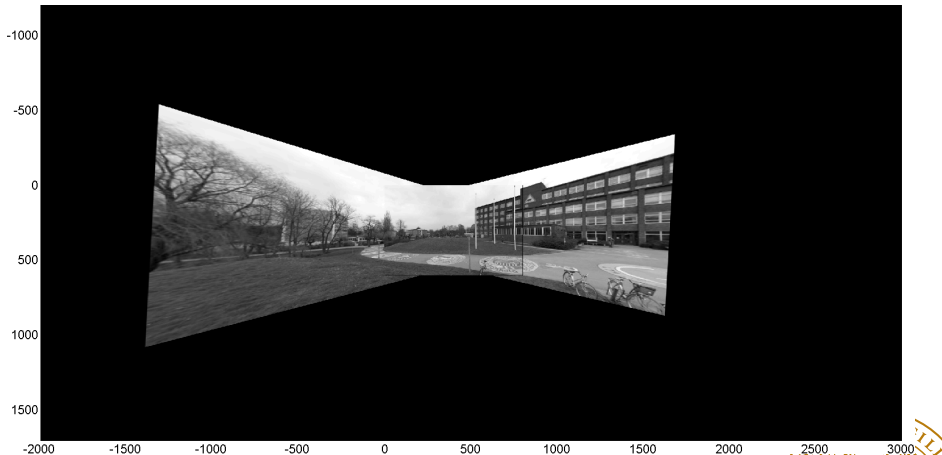
Panoramas



Points are transformed to the first image.

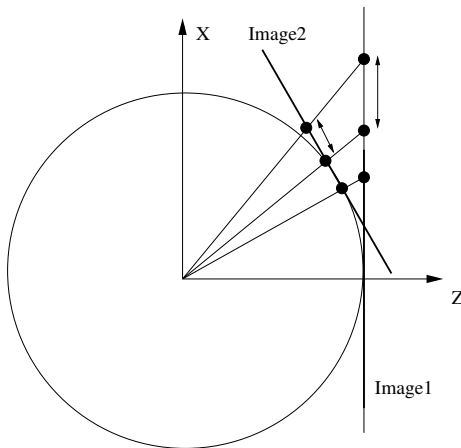


Panoramas



Panoramas

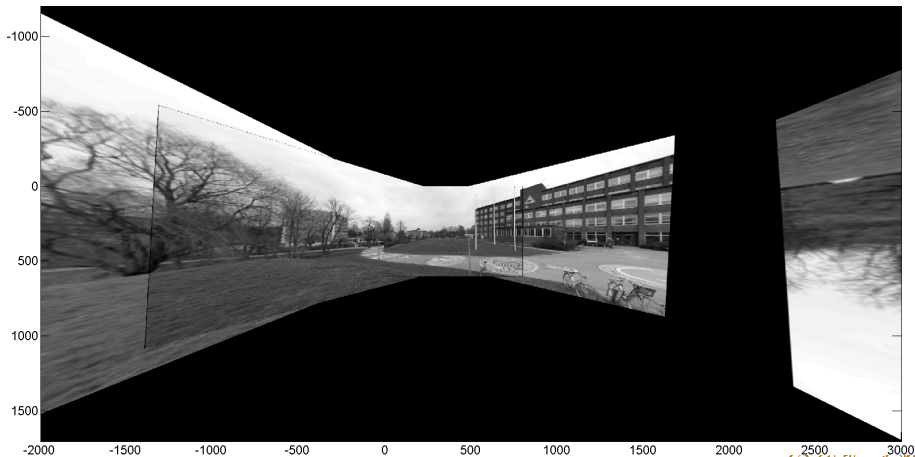
For calibrated cameras:



Distances are not preserved. Points close to the x -axis tend to infinity.

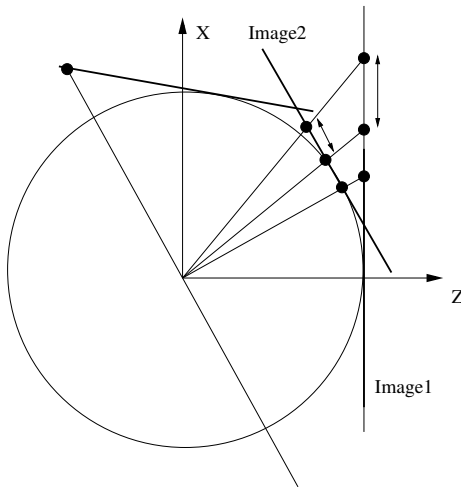


Panoramas



Panoramas

For calibrated cameras:

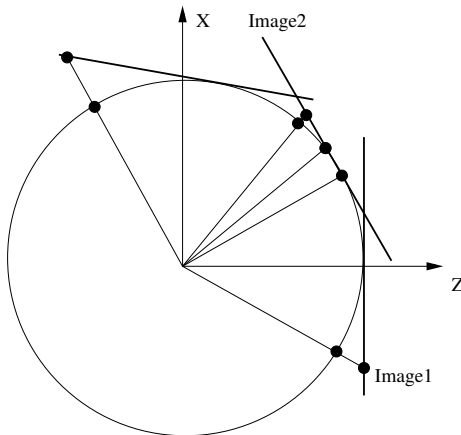


Cannot transfer all points into the first image.



Panoramas

For calibrated cameras:



Project onto a cylinder instead.



Panoramas

For calibrated cameras:



Distances are roughly preserved. Lines may not appear straight.



To do

- Work on assignment 2

