

CSCI 4830 / 5722

Computer Vision



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Computer Vision



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Spring 2019
Lecture 7



University of Colorado **Boulder**

Reminders

Submissions:

- Homework 2: due Wed 2/13 at 11 pm
- Homework 3: later this week

Readings:

- Szeliski:
 - chapter 3 (filters, changing resolution, Laplacian pyramids, warping)
 - chapter 4.1 (points) and 4.2 (edge detection)
- P&F Ch. 4,5
- Camera Calibration paper



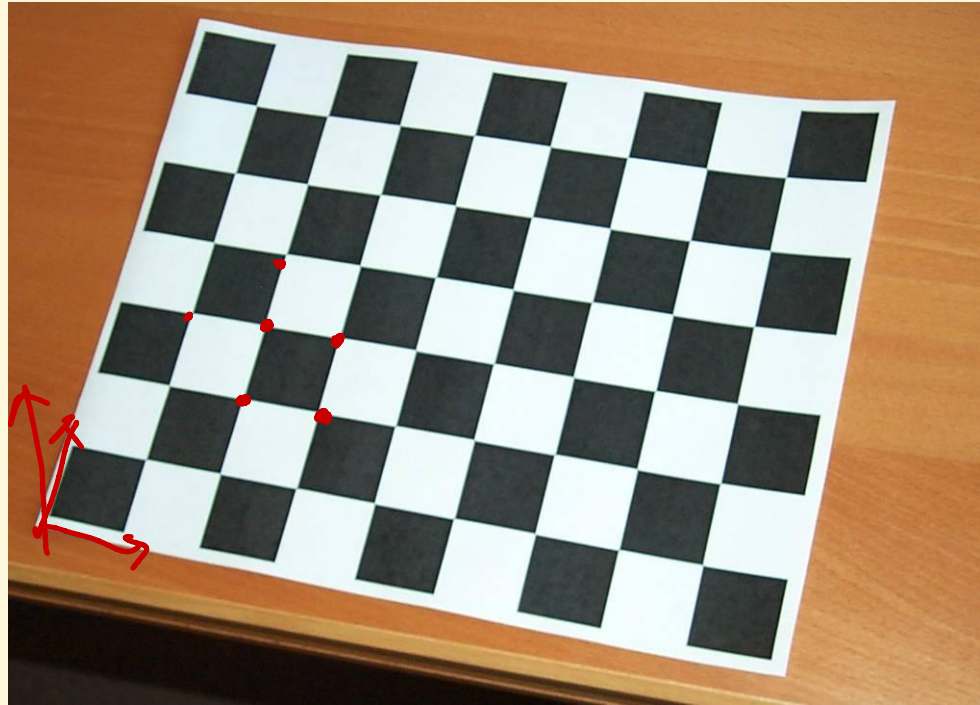
Today

- Camera Calibration paper – cont'
- Estimating homographies
- Homework 2



Camera Calibration: Problem Statement

Given one or more images of a calibration pattern, estimate the camera intrinsic (4 or more) and extrinsic parameters (6) using only observed camera data.



Camera calibration - II

- General strategy:
 - view calibration object; identify image points in image
 - positions are known in some fixed world coordinate system
 - optimization process: *discrepancy between observed image features and their theoretical position is minimized with respect to the camera's intrinsic and extrinsic parameters*
 - obtain camera matrix by minimizing error
 - obtain intrinsic parameters from camera matrix



Camera Calibration

- General strategy:
 - view calibration object
 - identify image points
 - obtain camera matrix by minimizing error
 - obtain intrinsic parameters from camera matrix
- Most modern systems employ the multi-plane method
 - avoids knowing absolute coordinates of calibration points

Methods:

- Photogrammetric Calibration
- Self Calibration
- **Multi-Plane Calibration**

Think of it like of a least squares problem (error minimization):

$$\begin{pmatrix} U \\ V \\ W \end{pmatrix}_{pix} = \begin{pmatrix} \alpha & \gamma & u_0 \\ 0 & \beta & v_0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} R & T \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} X \\ Y \\ Z \\ T \end{pmatrix}_{world}$$



Multi-Plane Calibration

- Hybrid method: Photogrammetric & Self-Calibration.
- Planar pattern imaged multiple times (inexpensive).
- Used widely in practice; many implementations.
- Based on a group of projective transformations called *homographies*.



Notation

image points

$$\tilde{\mathbf{m}} = [u, v, 1]^T \text{ and } \tilde{\mathbf{M}} = [X, Y, Z, 1]^T$$

world

$$s\tilde{\mathbf{m}} = A[R \quad T]\tilde{\mathbf{M}}$$

$$\mathbf{A} = \begin{bmatrix} \alpha & \gamma & u_0 \\ 0 & \beta & v_0 \\ 0 & 0 & 1 \end{bmatrix}$$

- s = arbitrary scale factor
- (u_0, v_0) = coordinates of the principal point
- α & β = scale factors in the u and v axes
- γ = skew factor between the two axes



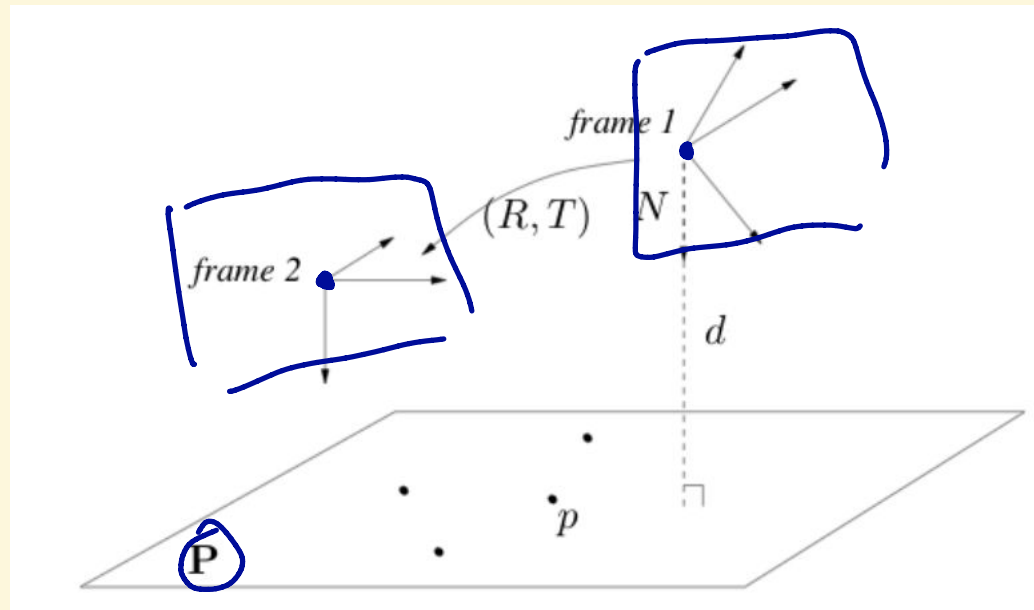
Planar Homographies

First Fundamental Theorem of Projective Geometry: There exists a unique homography H that performs a change of basis between two projective spaces of the same dimension.

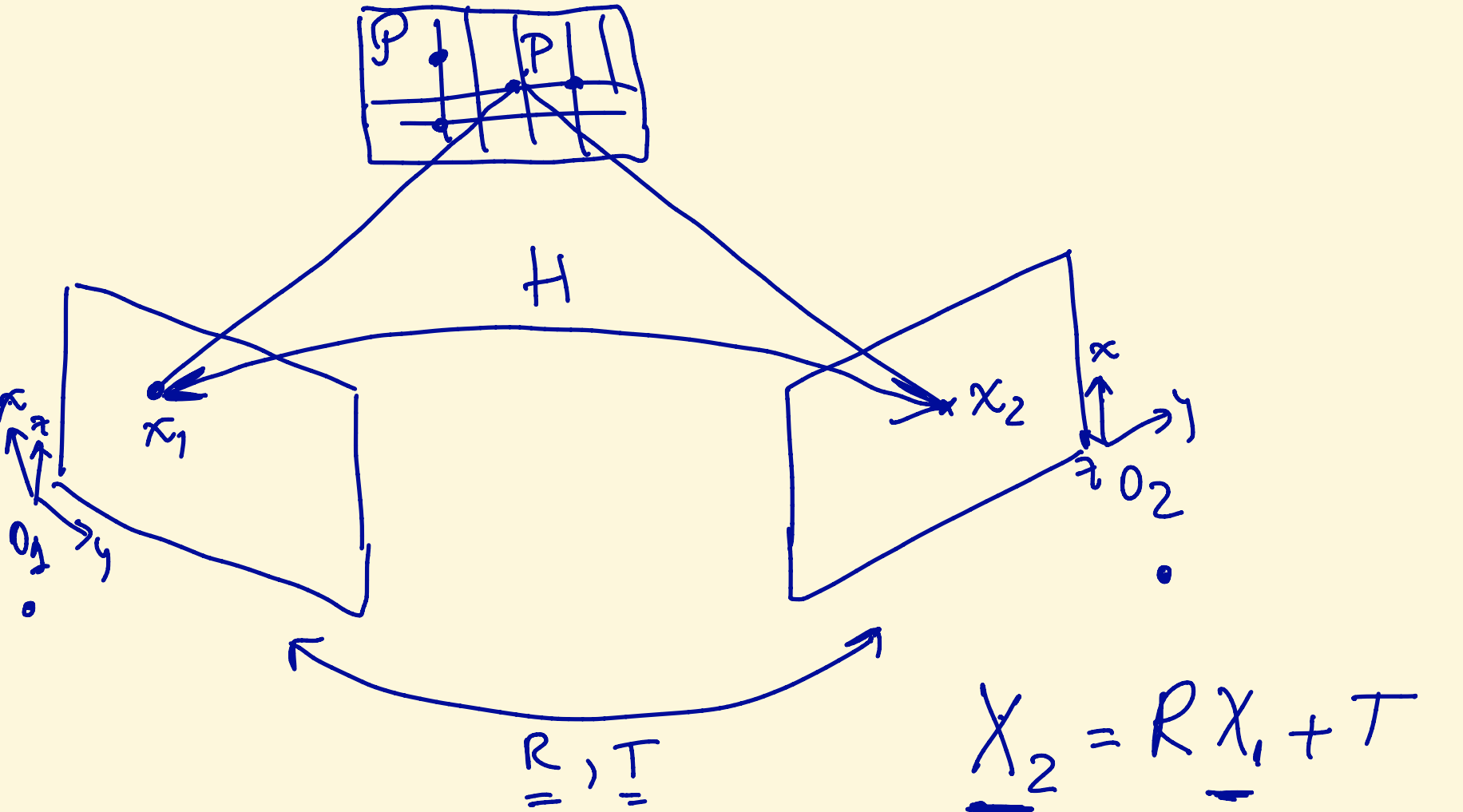
$$s\tilde{m} = H\tilde{M}$$



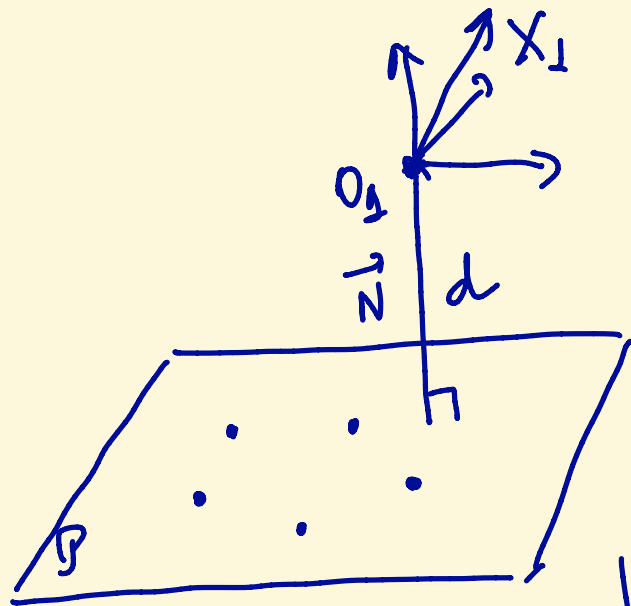
If two cameras image planar points from the same scene, then those cameras are related through a *homography*.



Planar Homographies



Planar Homographies



d - dist between 1st camera O and the plane

$N \rightarrow$ unit normal to plane P

$\vec{X}_1 \rightarrow$ a vector x_1

$$X_1 = \begin{bmatrix} x_1 \\ y_1 \\ z_1 \end{bmatrix}$$

$$N = \begin{bmatrix} a \\ b \\ c \end{bmatrix}$$

$$N^T X_1 = d$$

$$\frac{N^T X_1}{d} = 1$$

$$ax_1 + by_1 + cz_1 = d$$

$$X_2 = R X_1 + T = R X_1 + T * 1 = R X_1 + T \frac{N^T X_1}{d} =$$

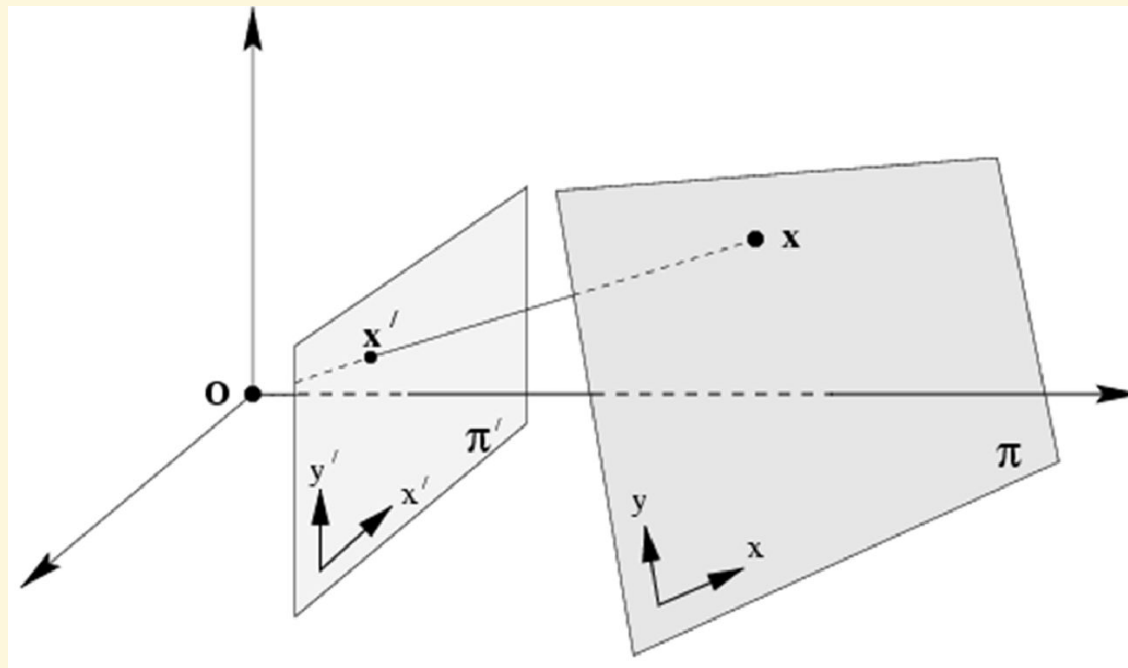
$$= \left(R + T \frac{N^T}{d} \right) X_1$$

$$H = \begin{bmatrix} R & T \\ N^T & d \end{bmatrix} \begin{matrix} 3 \times 3 \\ 3 \times 1 \\ 1 \times 3 \\ 1 \times 1 \end{matrix}$$

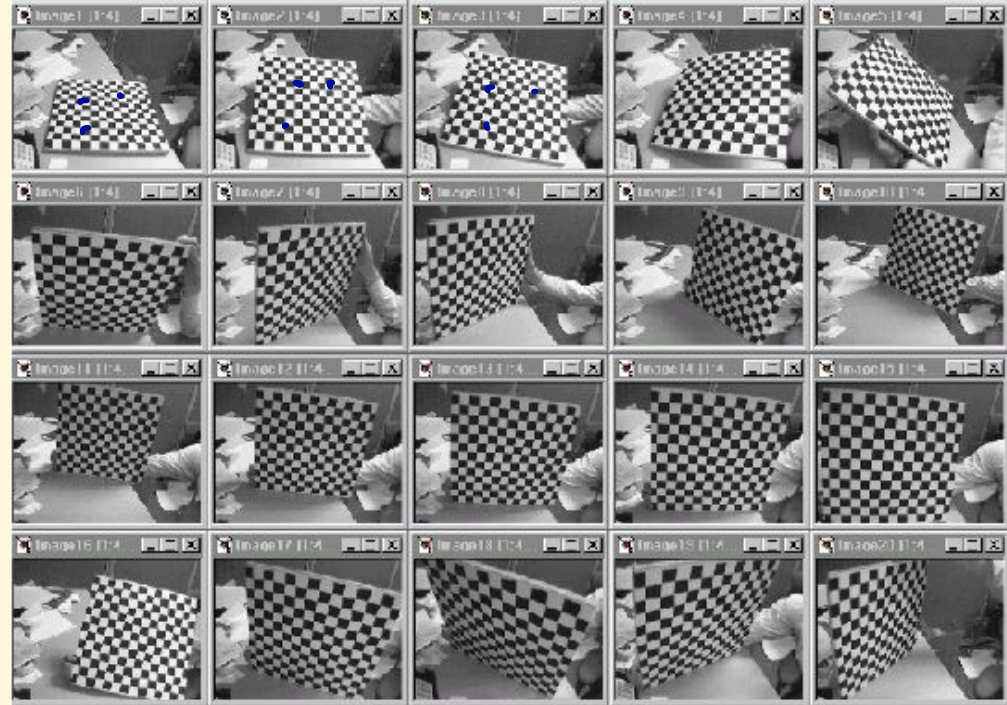
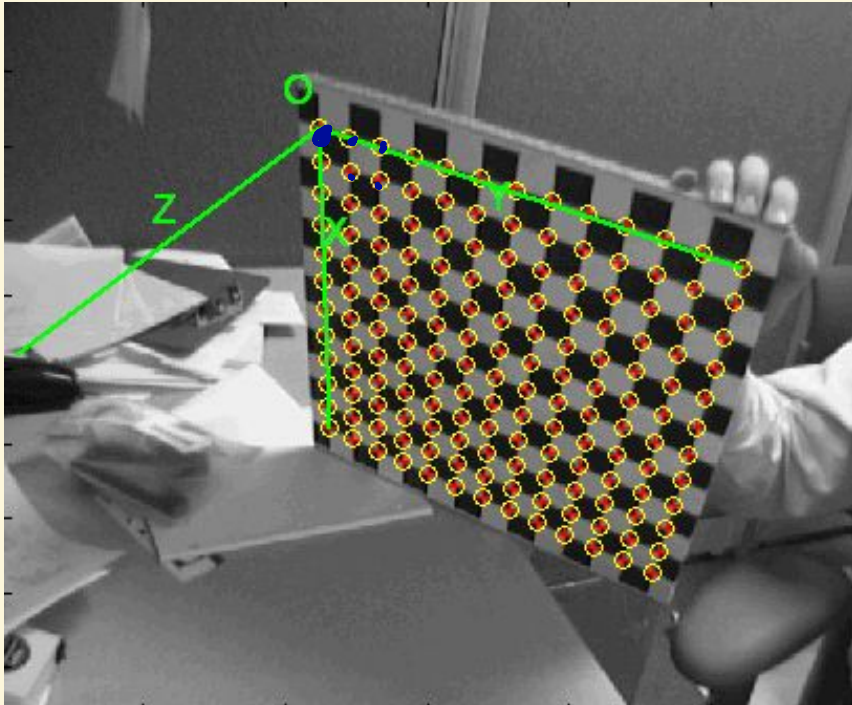


Planar Homographies

- Homography = transformation between 2 spaces of the same dimension
- In this case, 2 planes



Multi-Plane Calibration



Trick: set the world coordinate system to the corner of the checkerboard



Planar Homographies

$$m = A[R^T]M$$

$$\begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = A \begin{bmatrix} R & T \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} =$$

↑
intrinsics

$$\begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = A \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_x \\ r_{21} & r_{22} & r_{23} & t_y \\ r_{31} & r_{32} & r_{33} & t_z \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

→ 0

$$\begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = A \begin{bmatrix} r_{11} & r_{12} & t_x \\ r_{21} & r_{22} & t_y \\ r_{31} & r_{32} & t_z \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

$$\begin{pmatrix} 0, 5 \\ 0, 10 \end{pmatrix} \rightarrow \begin{pmatrix} 5, 0 \\ 5, 5 \end{pmatrix}$$

scale

$$sm = HM$$



Planar Homographies

$$s[u \ v \ 1]^T = A[r_1 \ r_2 \ r_3 \ t][X \ Y \ Z \ 1]^T$$

$$s[u \ v \ 1]^T = A[r_1 \ r_2 \ r_3 \ t][X \ Y \ 0 \ 1]^T$$

$$s[u \ v \ 1]^T = A[r_1 \ r_2 \ t][X \ Y \ 1]^T$$

$$s[u \ v \ 1]^T = H[X \ Y \ 1]^T$$

$$s\tilde{m} = H\tilde{M}$$



Planar Homographies

- Estimating a homography

$$m = \begin{bmatrix} u \\ v \\ 1 \end{bmatrix}$$

$$m = \lambda H M$$

$$\underline{m \otimes m} = \lambda m \otimes H M$$

$$0 = m \otimes H M$$

$$\hat{m}^T H M = 0$$

$$\begin{bmatrix} 0 & \textcircled{-1} & v \\ 1 & 0 & -u \\ -v & u & 0 \end{bmatrix} \begin{bmatrix} h_1^T M \\ \textcircled{h_2^T M} \\ h_3^T M \end{bmatrix} = 0$$

$$\begin{bmatrix} 0 & -h_2^T M + v h_3^T M = 0 \\ h_1^T M + 0 & -u h_3^T M = 0 \end{bmatrix}$$

$$\begin{bmatrix} 0^T & M^T & -v M^T \\ M^T & 0^T & -u M^T \end{bmatrix} \begin{bmatrix} h_1 \\ h_2 \\ h_3 \end{bmatrix} = 0$$

$$a \otimes b = \hat{a} * b \rightarrow \hat{a} \rightarrow \text{skew-symmetric matrix}$$

$$\hat{a} = \begin{bmatrix} 0 & -a_z & a_y \\ a_z & 0 & -a_x \\ -a_y & a_x & 0 \end{bmatrix}$$

rank 2

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

$$-h_1 = [h_{11} \ h_{12} \ h_{13}]$$

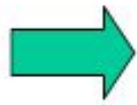
$$-h_2 = \dots$$

$$-h_3 = \dots$$

$$h_1^T M = M^T h_1$$



Estimating a Homography



$$\begin{aligned}x_2 &= \lambda H x_1 \\x_2 \times x_2 &= \lambda x_2 \times H x_1 \\ \hat{x}_2 H x_1 &= 0 \quad \text{or} \quad \lambda = 0\end{aligned}$$

Now $H = \begin{bmatrix} h_1^\top \\ h_2^\top \\ h_3^\top \end{bmatrix} \in \mathbb{R}^{3 \times 3}, h_i \in \mathbb{R}^3$. Hence $H x_1 = \begin{bmatrix} h_1^\top x_1 \\ h_2^\top x_1 \\ h_3^\top x_1 \end{bmatrix} \in \mathbb{R}^{3 \times 1}$. Let $x_2 = \begin{bmatrix} x \\ y \\ w \end{bmatrix}$. Then,

$$\hat{x}_2 = \begin{bmatrix} 0 & -w & y \\ w & 0 & -x \\ -y & x & 0 \end{bmatrix}.$$

$$\hat{x}_2 H x_1 = \begin{bmatrix} 0^\top & -w x_1^\top & y x_1^\top \\ w x_1^\top & 0^\top & -x x_1^\top \\ -y x_1^\top & x x_1^\top & 0^\top \end{bmatrix} \begin{bmatrix} h_1 \\ h_2 \\ h_3 \end{bmatrix} = 0 \quad \Rightarrow \quad \begin{bmatrix} A_1 \\ A_2 \\ A_3 \\ A_4 \end{bmatrix} h = 0$$



Homework 2 - mosaics

