

CZECH TECHNICAL UNIVERSITY IN PRAGUE

## 3D Computer Vision - Task 0-4 notes Lab session materials for subjects B4M33TDV, BE4M33TDV, XP33VID

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PERCEPTION

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Transformation between two projective planes

vector form elementwise form  $\lambda \underline{\mathbf{x}}_{2} = \mathbf{H} \underline{\mathbf{x}}_{1} \qquad \lambda \begin{bmatrix} u_{2} \\ v_{2} \\ w_{2} \end{bmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{bmatrix} u_{1} \\ v_{1} \\ w_{1} \end{bmatrix}$ 

- bijection, H is regular, invertible
- ▶  $\lambda \neq 0$ , any nonzero multiple of **H** represents the same transformation
- works for ideal points  $(w_1 = 0, w_2 = 0)$  as well
- in the affine plane (no ideal points) we can work with  $w_1 = 1$ ,  $w_2 = 1$  fixed

Notation of rows of  ${\bf H}$ 

$$\mathbf{H} = \left[ \begin{array}{c} \mathbf{h}_1^\top \\ \mathbf{h}_2^\top \\ \mathbf{h}_3^\top \end{array} \right]$$

i.e.,  $\mathbf{h}_i$  (without the  $^ op$ ) is the i-th row of the matrix transposed to the column vector.



Express particular elements of the vector <u>x</u><sub>2</sub>

 vector form
 elementwise form

  $\lambda u_2 = \mathbf{h}_1^\top \mathbf{x}_1$  (1)
  $\lambda u_2 = h_{11}u_1 + h_{12}v_1 + h_{13}w_1$ 
 $\lambda v_2 = \mathbf{h}_2^\top \mathbf{x}_1$  (2)
  $\lambda v_2 = h_{21}u_1 + h_{22}v_1 + h_{23}w_1$ 
 $\lambda w_2 = \mathbf{h}_3^\top \mathbf{x}_1$  (3)
  $\lambda w_2 = h_{31}u_1 + h_{32}v_1 + h_{33}w_1$ 

Eliminate nonzero λ – multiply sides of eq. (1) and (2) by swapped sides of eq. (3) divided by λ, i.e., multiply the right side of (1) and (2) by w<sub>2</sub> and the left side of (1) and (2) by h<sub>3</sub><sup>-</sup> x<sub>1</sub>/λ. Note: we do not divide by any of the coordinate entries (some can be zero).

$$\begin{split} & u_2 \mathbf{h}_3^\top \underline{\mathbf{x}}_1 = w_2 \mathbf{h}_1^\top \underline{\mathbf{x}}_1 \quad u_2(h_{31}u_1 + h_{32}v_1 + h_{33}w_1) = w_2(h_{11}u_1 + h_{12}v_1 + h_{13}w_1) \\ & v_2 \mathbf{h}_3^\top \underline{\mathbf{x}}_1 = w_2 \mathbf{h}_2^\top \underline{\mathbf{x}}_1 \quad v_2(h_{31}u_1 + h_{32}v_1 + h_{33}w_1) = w_2(h_{21}u_1 + h_{22}v_1 + h_{23}w_1) \end{split}$$

Some manipulation – to the homogeneous form, 'transpose' dot products

$$w_2 \mathbf{\underline{x}}_1^\top \mathbf{h}_1 \qquad -u_2 \mathbf{\underline{x}}_1^\top \mathbf{h}_3 = 0 \\ w_2 \mathbf{\underline{x}}_1^\top \mathbf{h}_2 \qquad -v_2 \mathbf{\underline{x}}_1^\top \mathbf{h}_3 = 0$$

 $\begin{array}{ll} w_2 u_1 h_{11} + w_2 v_1 h_{12} + w_2 w_1 h_{13} & -u_2 u_1 h_{31} - u_2 v_1 h_{32} - u_2 w_1 h_{33} = 0 \\ w_2 u_1 h_{21} + w_2 v_1 h_{22} + w_2 w_1 h_{23} & -v_2 u_1 h_{31} - v_2 v_1 h_{32} - v_2 w_1 h_{33} = 0 \end{array}$ 



Matrix representation – collect all known terms (point coordinates from the *i*-th corresponding pair of points) to a matrix A<sub>i</sub> and all unknowns to a vector h

$$\underbrace{\begin{bmatrix} w_{2}\mathbf{x}_{1}^{\top} & \mathbf{0}^{\top} & -u_{2}\mathbf{x}_{1}^{\top} \\ \mathbf{0}^{\top} & w_{2}\mathbf{x}_{1}^{\top} & -v_{2}\mathbf{x}_{1}^{\top} \end{bmatrix}}_{\mathbf{A}_{i}} \underbrace{\begin{bmatrix} \mathbf{h}_{1}^{\top} \\ \mathbf{h}_{2}^{\top} \\ \mathbf{h}_{3}^{\top} \end{bmatrix}}_{\mathbf{h}} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\underbrace{\begin{bmatrix} w_{2}u_{1} & w_{2}v_{1} & w_{2}w_{1} & 0 & 0 & 0 & -u_{2}u_{1} & -u_{2}v_{1} & -u_{2}w_{1} \\ 0 & 0 & 0 & w_{2}u_{1} & w_{2}v_{1} & w_{2}w_{1} & -v_{2}u_{1} & -v_{2}v_{1} & -v_{2}w_{1} \end{bmatrix}}_{\mathbf{A}_{i}} \underbrace{\begin{bmatrix} h_{11} \\ h_{12} \\ h_{13} \\ h_{22} \\ h_{23} \\ h_{31} \\ h_{32} \\ h_{33} \\ h_{32} \\ h_{33} \\ h_{33} \\ h_{32} \\ h_{33} \\ h_{33} \\ h_{32} \\ h_{33} \\ h_{32} \\ h_{33} \\ h_{32} \\ h_{33} \\ h_{33} \\ h_{32} \\ h_{33} \\ h_{33} \\ h_{32} \\ h_{33} \\ h_{33} \\ h_{33} \\ h_{32} \\ h_{33} \\$$

▶  $A_i - 2$  equations, h - 8 unknowns (up to scalar multiple)  $\rightarrow 4$  correspondences needed, stacked to  $8 \times 9$  matrix A. Then h is a solution of linear homogeneous system, i.e. right null-space of A, and H is composed from h.

$$\mathbf{A} = \begin{bmatrix} \vdots \\ \mathbf{A}_{\mathbf{i}} \\ \vdots \end{bmatrix} \qquad \mathbf{A}\mathbf{h} = \mathbf{0}$$



Note: When working with points from the affine plane only, we can assume point coordinates normalized to  $w_1 = 1, w_2 = 1$ .

$$\mathbf{A} = \begin{bmatrix} \vdots \\ \mathbf{x}_{1}^{\mathsf{T}} & \mathbf{0}^{\mathsf{T}} & -u_{2}\mathbf{x}_{1}^{\mathsf{T}} \\ \mathbf{0}^{\mathsf{T}} & \mathbf{x}_{1}^{\mathsf{T}} & -v_{2}\mathbf{x}_{1}^{\mathsf{T}} \end{bmatrix}$$
$$\mathbf{A} = \begin{bmatrix} u_{1} & v_{1} & 1 & 0 & 0 & 0 & -u_{2}u_{1} & -u_{2}v_{1} & -u_{2} \\ u_{1} & v_{1} & 1 & 0 & 0 & 0 & -u_{2}u_{1} & -u_{2}v_{1} & -u_{2} \\ 0 & 0 & 0 & u_{1} & v_{1} & 1 & -v_{2}u_{1} & -v_{2}v_{1} & -v_{2} \\ \vdots & \vdots & & \vdots & & \end{bmatrix}$$



- Homography between two images does not depends on world coordinate system choice (chosen at the first camera matrix P1)
- Two cameras  $\mathbf{P}_1$  and  $\mathbf{P}_2$  observing a plane  $\mathbf{\Pi}$

$$\mathbf{P}_1 = \mathbf{K}_1 \begin{bmatrix} \mathbf{I} & \mathbf{0} \end{bmatrix} \quad \mathbf{P}_2 = \mathbf{K}_2 \begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix} \quad \underline{\mathbf{\Pi}}^\top = \begin{bmatrix} \mathbf{n}^\top & d \end{bmatrix}$$

1. Reconstruct  $\underline{\mathbf{X}}$  constrained by the plane  $\underline{\mathbf{\Pi}}$  from  $\underline{\mathbf{u}}_1$ (projection equation augmented by plane constraint row  $\underline{\mathbf{\Pi}}^{\top} \underline{\mathbf{X}} = 0$  to obtain  $4 \times 4$  invertible matrix

$$\lambda_{1} \begin{bmatrix} \mathbf{\underline{u}}_{1} \\ 0 \end{bmatrix} = \underbrace{\begin{bmatrix} \mathbf{K}_{1} & \mathbf{0} \\ \mathbf{n}^{\top} & d \end{bmatrix}}_{\mathbf{M}} \mathbf{\underline{X}}$$
$$\mathbf{\underline{X}} = \lambda_{1} \underbrace{\begin{bmatrix} \mathbf{K}_{1}^{-1} & \mathbf{0} \\ \frac{-\mathbf{n}^{\top}\mathbf{K}_{1}^{-1}}{d} & \frac{1}{d} \end{bmatrix}}_{\mathbf{M}-1} \begin{bmatrix} \mathbf{\underline{u}}_{1} \\ 0 \end{bmatrix} = \lambda_{1} \begin{bmatrix} \mathbf{K}_{1}^{-1} \\ \frac{-\mathbf{n}^{\top}\mathbf{K}_{1}^{-1}}{d} \end{bmatrix} \mathbf{\underline{u}}_{1}$$

2. Project  $\underline{\mathbf{X}}$  to the second camera to obtain  $\mathbf{H}$ 

$$\lambda_{2}\underline{\mathbf{u}}_{2} = \mathbf{P}_{2}\underline{\mathbf{X}} = \lambda_{1}\mathbf{K}_{2}\begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix} \begin{bmatrix} \mathbf{K}_{1}^{-1} \\ \frac{-\mathbf{n}^{\top}\mathbf{K}_{1}^{-1}}{d} \end{bmatrix} \underline{\mathbf{u}}_{1} = \lambda_{1}\underbrace{\mathbf{K}_{2}\left(\mathbf{R} - \mathbf{t}\mathbf{n}^{\top}/d\right)\mathbf{K}_{1}^{-1}}_{\mathbf{H}_{12}} \underline{\mathbf{u}}_{1}$$



General cameras – derivation is the same, but a bit ugly

$$\lambda_1 \begin{bmatrix} \mathbf{\underline{u}} \\ 0 \end{bmatrix} = \begin{bmatrix} \mathbf{K_1} \mathbf{R_1} & -\mathbf{K_1} \mathbf{R_1} \mathbf{C_1} \\ \mathbf{n}^\top & d \end{bmatrix} \mathbf{\underline{X}} \to \mathbf{\underline{X}} = \lambda_1 \begin{bmatrix} \mathbf{R}_1^\top \mathbf{K}_1^{-1} - \frac{\mathbf{C_1} \mathbf{n}^\top \mathbf{R}_1^\top \mathbf{K}_1^{-1}}{d + \mathbf{n}^\top \mathbf{C_1}} \\ \frac{-\mathbf{n}^\top \mathbf{R}_1^\top \mathbf{K}_1^{-1}}{d + \mathbf{n}^\top \mathbf{C_1}} \end{bmatrix} \mathbf{\underline{u}}$$

$$\lambda_2 \underline{\mathbf{u}}_2 = \mathbf{P}_2 \underline{\mathbf{X}} = \lambda_1 \mathbf{K}_2 \begin{bmatrix} \mathbf{R}_2 & -\mathbf{R}_2 \mathbf{C}_2 \end{bmatrix} \begin{bmatrix} \mathbf{R}_1^\top - \frac{\mathbf{C}_1 \mathbf{n}^\top \mathbf{R}_1^\top}{d + \mathbf{n}^\top \mathbf{C}_1} \\ \frac{-\mathbf{n}^\top \mathbf{R}_1^\top}{d + \mathbf{n}^\top \mathbf{C}_1} \end{bmatrix} \mathbf{K}_1^{-1} \underline{\mathbf{u}}$$

$$\mathbf{H} = \mathbf{K}_2 \left( \mathbf{R}_2 \mathbf{R}_1^\top - \frac{\mathbf{R}_2 (\mathbf{C}_1 - \mathbf{C}_2) \mathbf{n}^\top \mathbf{R}_1^\top}{d + \mathbf{n}^\top \mathbf{C}_1} \right) \mathbf{K}_1^{-1}$$

Inverse homography

1. Change of coordinate frame - apply T (4 × 4):  $\underline{\mathbf{X}}' = \mathbf{T}\underline{\mathbf{X}}, \ \mathbf{P}' = \mathbf{P}\mathbf{T}^{-1}, \ \underline{\mathbf{\Pi}}'^{\top} = \underline{\mathbf{\Pi}}\mathbf{T}^{-1}$ 

$$\mathbf{T}^{-1} = \left[ \begin{array}{cc} \mathbf{R}^\top & -\mathbf{R}^\top \mathbf{t} \\ \mathbf{0}^\top & 1 \end{array} \right]$$

$$\mathbf{P}_1' = \mathbf{K}_1 \begin{bmatrix} \mathbf{R}^\top & -\mathbf{R}^\top \mathbf{t} \end{bmatrix} \quad \mathbf{P}_2' = \mathbf{K}_2 \begin{bmatrix} \mathbf{I} & \mathbf{0} \end{bmatrix} \quad \underline{\Pi}'^\top = \begin{bmatrix} \mathbf{n}^\top \mathbf{R}^\top & -\mathbf{n}^\top \mathbf{R}^\top \mathbf{t} + d \\ \mathbf{n}'^\top & \mathbf{n}'^\top \end{bmatrix}$$

 $\mathbf{n}', d'$  are parameters of the plane w.r.t. coordinate system of the second camera 2. Use eq. from Page 6, substitute  $\mathbf{P}_1 \leftarrow \mathbf{P}'_2$ ,  $\mathbf{P}_2 \leftarrow \mathbf{P}'_1$ ,  $\underline{\Pi} \leftarrow \underline{\Pi}'$ 

$$\mathbf{H}_{21} = \mathbf{K}_1 \left( \mathbf{R}^\top + \frac{\mathbf{R}^\top \mathbf{t} \mathbf{n}^\top \mathbf{R}^\top}{d - \mathbf{n}^\top \mathbf{R}^\top \mathbf{t}} \right) \mathbf{K}_2^{-1} = \mathbf{K}_1 \left( \mathbf{R}^\top + \frac{\mathbf{R}^\top \mathbf{t} \mathbf{n}^{\prime\top}}{d'} \right) \mathbf{K}_2^{-1}$$

(can be easily verified that  $\mathbf{H}_{21}=\mathbf{H}_{12}^{-1}$  from Page 6)

Alternative: use the general equation on Page 7 for  ${f P}_1,\,{f P}_2$  swapped and the same  ${f \underline{\Pi}}.$ 



- A pair of cameras  $\mathbf{K_1}[\mathbf{I}|\mathbf{0}]$ ,  $\mathbf{K_2}[\mathbf{R}|\mathbf{t}]$  observes a pair of planes  $\underline{\Pi}_a$ ,  $\underline{\Pi}_b \rightarrow \mathbf{H}_{12}^a$ ,  $\mathbf{H}_{12}^b$
- ▶ Consider a composed homography 'there-and-back':  $1 \rightarrow 2$  via  $\underline{\Pi}_a$  and  $2 \rightarrow 1$  via  $\underline{\Pi}_b$

$$\mathbf{H} = \mathbf{H}_{21}^{b} \mathbf{H}_{12}^{a} = \mathbf{K}_{1} (\mathbf{R}^{\top} + \mathbf{R}^{\top} \mathbf{t} \mathbf{n}_{b}^{\prime \top} / d_{b}^{\prime}) \mathbf{K}_{2}^{-1} \mathbf{K}_{2} (\mathbf{R} - \mathbf{t} \mathbf{n}_{a}^{\top} / d_{a}) \mathbf{K}_{1}^{-1} =$$

$$= \mathbf{I} + \underbrace{(-\mathbf{K}_{1}\mathbf{R}^{\top}\mathbf{t})}_{\mathbf{v}} \underbrace{\left(\frac{\mathbf{n}_{a}^{\top}}{d_{a}} - \frac{\mathbf{n}_{b}^{\prime\top}\mathbf{R}}{d_{b}^{\prime}} + \frac{\mathbf{n}_{b}^{\prime\top}\mathbf{t}\mathbf{n}_{a}^{\top}}{d_{a}d_{b}^{\prime}}\right)\mathbf{K}_{1}^{-1}}_{\mathbf{a}^{\top}} = \mathbf{I} + \mathbf{v}\mathbf{a}^{\top}$$

Planar homology H with vertex v, axis a, eigen value  $\mu = 1 + \mathbf{a}^{\top} \mathbf{v}$ :  $\mathbf{H} = \mathbf{I} + \mathbf{v} \mathbf{a}^{\top}$   $\mathbf{H} \mathbf{v} = \mathbf{v} + \mathbf{v} \mathbf{a}^{\top} \mathbf{v} = (1 + \mathbf{a}^{\top} \mathbf{v}) \mathbf{v} = \mu \mathbf{v}$  v is eigenvector of H corresponding to eigenvalue  $\mu$ , as a planar point it is fixed w.r.t. the transformation  $\mathbf{a}^{\top} \mathbf{x} = 0 \Longrightarrow \mathbf{H} \mathbf{x} = \mathbf{x} + \mathbf{v} \mathbf{a}^{\top} \mathbf{x} = \mathbf{x}$  a is a line of fixed points - 2D eigenspace of H with double eigenvalue 1

- $\blacktriangleright$  a and  ${\bf v}$  represents homogeneous image entities but their scale matters in  ${\bf I}+{\bf va}^\top$
- ▶ the point **v** and all points on a line **a** in one image are mapped to the second image to a points same for both  $\mathbf{H}_{12}^a$  and  $\mathbf{H}_{12}^b$ :  $\mathbf{H}_{12}^a \mathbf{v} \simeq \mathbf{H}_{12}^b \mathbf{v}$   $(\mathbf{H}_{12}^a)^{-\top} \mathbf{a} \simeq (\mathbf{H}_{12}^b)^{-\top} \mathbf{a}$
- a is the image of the common line of the planes
- $\mathbf{v} = -\mathbf{K_1}\mathbf{R}^{\top}\mathbf{t}$  is the epipole in the first image
- eigenvectors of **H** are  $(\mathbf{v}, \mathbf{x}_1, \mathbf{x}_2)$  corresponding to eigenvalues  $(\mu, 1, 1)$ , and  $\mathbf{a} = \mathbf{x}_1 \times \mathbf{x}_2$ (when the matrix is multiplied by an unknown scale  $\lambda$ , the eigenvalues become  $(\lambda \mu, \lambda, \lambda)$ )



1. A point  $\underline{\mathbf{u}}$  and its image  $\lambda \underline{\mathbf{u}}' = \mathbf{H} \underline{\mathbf{u}}$  define a common line  $\underline{\mathbf{k}}$  (if not lying on  $\mathbf{a}$  or  $\mathbf{v}$ ). Since  $\mathbf{a}^{\top} \underline{\mathbf{u}} \neq \mathbf{0}$ , the vertex must lie on this line as well:

$$\underline{\mathbf{k}}^{\top} \, \underline{\mathbf{u}} = 0 \,, \quad \underline{\mathbf{k}}^{\top} \, \mathbf{H} \, \mathbf{u} = 0$$
$$\underline{\mathbf{k}}^{\top} \left( \underline{\mathbf{u}} + \mathbf{v} \mathbf{a}^{\top} \, \underline{\mathbf{u}} \right) = 0 \implies \underline{\mathbf{k}}^{\top} \, \mathbf{v} \mathbf{a}^{\top} \, \underline{\mathbf{u}} = 0 \implies \underline{\mathbf{k}}^{\top} \, \mathbf{v} = 0$$

The vertex is estimated from two correspondences:  $\mathbf{v} = (\underline{\mathbf{u}}_1 \times \underline{\mathbf{u}}'_1) \times (\underline{\mathbf{u}}_2 \times \underline{\mathbf{u}}'_2)$ 

2. When the vertex v is known, a linear non-homogeneous system can be used for computing a. Let  $\mathbf{H} = \mathbf{I} + \mathbf{va}^{\top}$ ,  $\mathbf{\underline{u}}_{i}^{\top} = [x_{i}, y_{i}, w_{i}]$ ,  $\mathbf{\underline{u}}_{i}^{\prime\top} = [x_{i}', y_{i}', w_{i}']$ ,  $\mathbf{v}^{\top} = [x_{v}, y_{v}, w_{v}]$ .

$$\boldsymbol{\lambda} \begin{bmatrix} x_i' \\ y_i' \\ w_i' \end{bmatrix} = \underline{\mathbf{u}}_i + \mathbf{v} \mathbf{a}^\top \underline{\mathbf{u}}_i = \underline{\mathbf{u}}_i + \mathbf{v} \underline{\mathbf{u}}_i^\top \mathbf{a} = \begin{bmatrix} x_i \\ y_i \\ w_i \end{bmatrix} + \begin{bmatrix} x_v \underline{\mathbf{u}}_i^\top \\ y_v \underline{\mathbf{u}}_i^\top \\ w_v \underline{\mathbf{u}}_i^\top \end{bmatrix} \mathbf{a}$$

The left sides of these eqs. consists of the same vector  $\underline{u}_i^{\top}$  multiplied by a scalar, i.e., they are linearly dependent, and only one eq. can be used  $\rightarrow$  three points needed.

$$\mathbf{A} = \begin{bmatrix} (x_1'w_v - w_1'x_v)\mathbf{\underline{u}}_1^\top \\ (x_2'w_v - w_2'x_v)\mathbf{\underline{u}}_2^\top \\ (x_3'w_v - w_3'x_v)\mathbf{\underline{u}}_3^\top \end{bmatrix} \quad \mathbf{b} = \begin{bmatrix} x_1w_1' - w_1x_1' \\ x_2w_2' - w_2x_2' \\ x_3w_3' - w_3x_3' \end{bmatrix} \quad \mathbf{a} = \mathbf{A}^{-1}\mathbf{b}$$

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