

# Geometric definition of homography

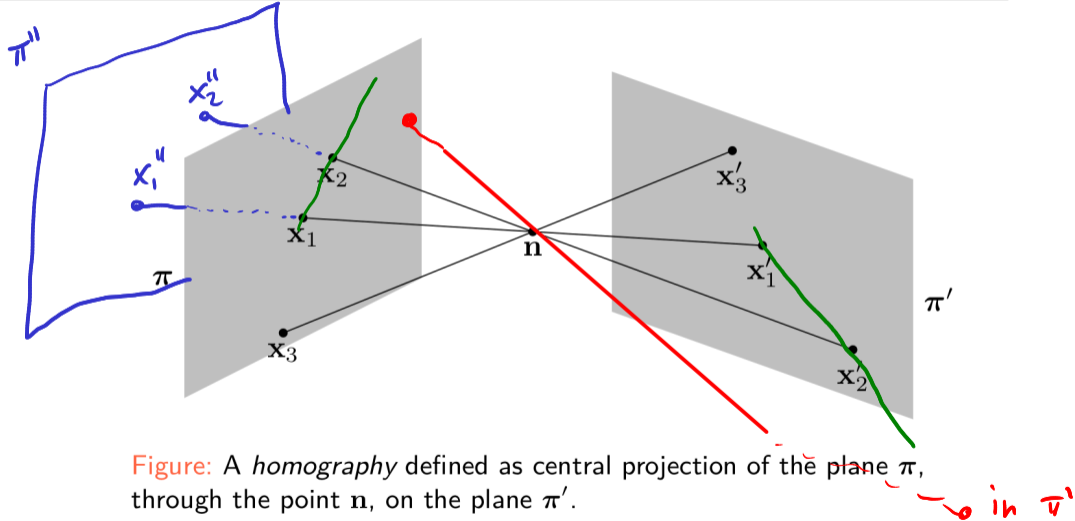


Figure: A homography defined as central projection of the plane  $\pi$ , through the point  $n$ , on the plane  $\pi'$ .

# Observations

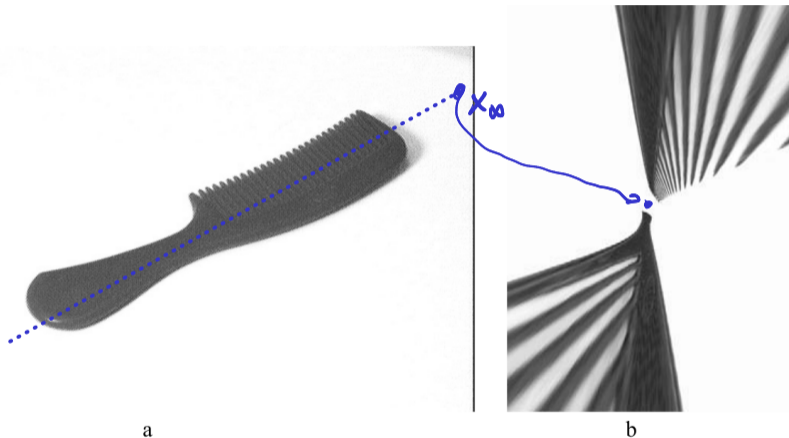
- A homography is always invertible.
- A homography becomes a perspective pinhole camera, if everything in the scene lies in a plane.
- The two planes can lie on the 'same side' of the projection centre.
- A homography maps straight lines in one image to straight lines in the other image.  
The "only" invariant of a homography!
- A homography can map proper points to ideal points, and vice versa.

# Example of a homography



**Figure:** The right image is a synthetic view generated from the left image by applying a suitable homography.

# Example of a homography (crazy case)



(From Hartley & Zisserman, "Multiple View Geometry in Computer Vision", 2004)

# Algebraic representation of homographies

A homography is represented by an invertible 3x3 matrix,

$$x' \sim H_1 x \Leftrightarrow x' \sim H_2 x$$

and

$$H_1 \sim H_2$$

all 3x3 invertible matrices represent homographies.

$\Rightarrow H_1$  and  $H_2$   
represent the  
same transform

Generalises affine transforms:

recall:

$$T_{\text{affine}} = \begin{pmatrix} A & t \\ 0^T & 1 \end{pmatrix} \left| \begin{array}{l} H = \begin{pmatrix} A & t \\ c^T & d \end{pmatrix} \\ \text{but must} \\ \text{be invertible!} \end{array} \right.$$

Invariants: Affine parallel lines / homographies straight lines

Degrees of Freedom: Not 9! Eight! (remove one for scaling)

# Homographies & Cameras

## Theorem

centre:  $P'c=0 \Rightarrow c \sim \begin{pmatrix} -A^{-1}b \\ 1 \end{pmatrix}$

Assume  $\mathbf{P} = [\mathbf{I} \ \mathbf{0}]$  and  $\mathbf{P}' = [\mathbf{A} \ \mathbf{b}]$  are projective cameras, and let  $\pi = (\nu, d)$  be a plane which does not contain any of the camera centres. If  $\mathbf{X}$  is a point on the plane and projects into the two views as  $\mathbf{x} \sim \mathbf{P}\mathbf{X}$  and  $\mathbf{x}' \sim \mathbf{P}'\mathbf{X}$ , then

$$\mathbf{x}' \sim (\mathbf{A} - \mathbf{b}\nu^\top / d)\mathbf{x},$$

and the matrix  $\mathbf{H} = \mathbf{A} - \mathbf{b}\nu^\top / d$  is a homography.

splitting up  $H$  into  $R, t, n$  is the "homography decomposition problem"

Assumption:  $P=K[I \ 0]$  and  $P'=K[R \ t]$

# Homographies & Cameras (contd.)

Proof (except invertibility):

$$\text{We know } x \sim P X = \begin{pmatrix} I & 0 \\ 0 & 1 \end{pmatrix} X = \begin{pmatrix} I & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ s \end{pmatrix}$$

$3 \times 4$     $4 \times 1$

$\tilde{X}(s)$

In the second camera:

$$x' \sim P' X(s) = (A \ b) \begin{pmatrix} x \\ s \end{pmatrix} = Ax + bs$$

Since  $X(s)$  lies on  $\Pi$ , we have  $\pi^T X(s) = 0 \Leftrightarrow (v^T \ d) \begin{pmatrix} x \\ s \end{pmatrix} = 0 \Leftrightarrow v^T x + ds = 0 \Rightarrow s = -\frac{v^T x}{d}$

note:  $d \neq 0$   
by design

$$\Rightarrow x' \sim Ax + bs = Ax - \frac{bv^T x}{d} = \left(A - \frac{bv^T}{d}\right) x \quad \therefore$$

# Cameras with identical centre

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Consider  $P=(I\ 0)$  and  $P'=(A\ 0)$ .



Exactly the situation  
on the first slide!

Generate panoramas!

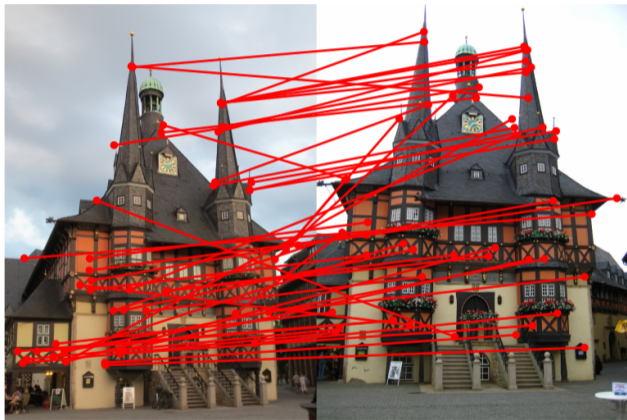


# Point Correspondences

In many cases it is useful to consider point correspondences between two (or more) images.

$$X_j \leftrightarrow X'_j, \quad j=1, \dots, n$$

- There exist many methods for automatically finding (putative) point correspondences, based on feature extraction (SIFT, SURF, ...).
- Not all correspondences they propose are correct! ("Outliers", more on that in TSBB15.)



# Direct Linear Transformation (DLT) $[u]_x v = u \times v$

Suppose we are given a number of point correspondences  $x_j \leftrightarrow x_j'$   
and want to find a homography that transforms all  $x_j$  into  $x_j'$ .

Equations: We want  $x_j' \sim Hx_j$  not an equation! Note:  $x_j' \in \mathbb{R}^3, Hx_j \in \mathbb{R}^3$

Formulate using cross product!  
 $\Rightarrow x_j' \times Hx_j = 0 \Leftrightarrow [x_j']_x Hx_j = 0$   
unknowns!  
known

Pull H outside using vec/Kronecker!

$$x_j = (u_j, v_j, 1), \quad x_j' = (u_j', v_j', 1)$$

$$A_j = \begin{pmatrix} u_j [x_j]_x & v_j [x_j]_x & [x_j]_x \end{pmatrix} = \begin{pmatrix} 0 & -u_j & u_j v_j' & 0 & -v_j & v_j v_j' & 0 & -1 & v_j \\ u_j & 0 & -u_j u_j' & v_j & 0 & -v_j u_j' & 1 & 0 & -u_j \\ -u_j v_j' & u_j u_j' & 0 & -v_j v_j' & v_j u_j' & 0 & -v_j & u_j & 0 \end{pmatrix}$$

$$\Rightarrow \underbrace{(x_j'^T \otimes [x_j]_x)}_{A_j} \text{vec } H = 0$$

DLT constraint

Conclusion: Cumbersome! Better to use Kronecker...

# Direct Linear Transformation (DLT)

(contd.)

We stack all DLT constraints for  $j=1, \dots, n$  into a "data matrix"  $A$ :

$$A_j = x_j^T \otimes [x_j^*]_x$$

$3 \times 9$

$$A = \begin{pmatrix} A_1 \\ \vdots \\ A_n \end{pmatrix}$$

$3n \times 9$

DLT system:  $A \text{ vec } H = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$   $3n$  zeros

Note:  $\text{rank } A_j = 2 \Rightarrow \text{rank } A \leq 2n$

But  $\text{rank } A \leq 9$

$\text{rank } A \leq \min(2n, 9)$

Four point correspondences gives "unique" solution  $H$