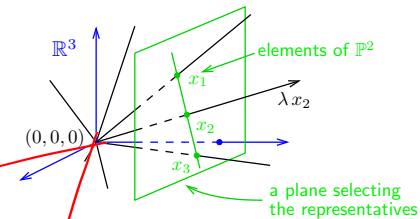


## ► Homography in $\mathbb{P}^2$



Projective plane  $\mathbb{P}^2$ : Vector space of dimension 3 excluding the zero vector,  $\mathbb{R}^3 \setminus (0, 0, 0)$ , factorized to linear equivalence classes ('rays')

including 'points at infinity'

**Homography in  $\mathbb{P}^2$ :** Non-singular linear mapping in  $\mathbb{P}^2$

$$\underline{\mathbf{x}}' \simeq \mathbf{H} \underline{\mathbf{x}}, \quad \mathbf{H} \in \mathbb{R}^{3,3} \text{ non-singular}$$

$$\lambda \mathbf{H} \simeq \mathbf{H} \quad \lambda \neq 0$$

an analogic definition for  $\mathbb{P}^3$



**Defining properties**

- collinear image points are mapped to collinear image points

lines of points are mapped to lines of points

- concurrent image lines are mapped to concurrent image lines

- and point-line incidence is preserved

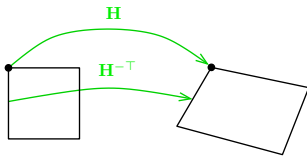


concurrent = intersecting at a point

e.g. line intersection points mapped to line intersection points

- $\mathbf{H}$  is a  $3 \times 3$  non-singular matrix,  $\lambda \mathbf{H} \simeq \mathbf{H}$  equivalence class, 8 degrees of freedom
- homogeneous matrix representant:  $\det \mathbf{H} = 1$
- what we call homography here is often called 'projective collineation' in mathematics

## ► Mapping 2D Points and Lines by Homography



$$\underline{\mathbf{m}}' \simeq \mathbf{H} \underline{\mathbf{m}} \quad \text{image point}$$

$$\underline{\mathbf{n}}' \simeq \mathbf{H}^{-T} \underline{\mathbf{n}} \quad \text{image line}$$

$$\mathbf{H}^{-T} = (\mathbf{H}^{-1})^T = (\mathbf{H}^T)^{-1}$$

- incidence is preserved:  $(\underline{\mathbf{m}}')^T \underline{\mathbf{n}}' \simeq \underline{\mathbf{m}}^T \mathbf{H}^T \mathbf{H}^{-T} \underline{\mathbf{n}} = \underline{\mathbf{m}}^T \underline{\mathbf{n}} = 0$

Mapping a finite 2D point  $\mathbf{m} = (u, v)$  to  $\underline{\mathbf{m}} = (u', v')$

- extend the Cartesian (pixel) coordinates to homogeneous coordinates,  $\underline{\mathbf{m}} = (u, v, 1)$
- map by homography,  $\underline{\mathbf{m}}' = \mathbf{H} \underline{\mathbf{m}}$
- if  $m'_3 \neq 0$  convert the result  $\underline{\mathbf{m}}' = (m'_1, m'_2, m'_3)$  back to Cartesian coordinates (pixels),

$$u' = \frac{m'_1}{m'_3}, \quad v' = \frac{m'_2}{m'_3}$$

- note that, typically,  $m'_3 \neq 1$
- an infinite point  $(u, v, 0)$  maps the same way

$m'_3 = 1$  when  $\mathbf{H}$  is affine

# Some Homographic Tasters

**Rectification of camera rotation:** →59 (geometry), →122 (homography estimation)



$$\mathbf{H} \simeq \mathbf{K}\mathbf{R}^T\mathbf{K}^{-1}$$

maps from image plane to facade plane

**Homographic Mouse for Visual Odometry:** [Mallis 2007]



illustrations courtesy of AMSL Racing Team, Meiji University and LIBVISO: Library for VISual Odometry

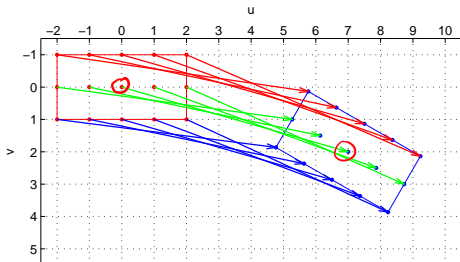
$$\mathbf{H} \simeq \mathbf{K} \left( \mathbf{R} - \frac{\mathbf{t}\mathbf{n}^T}{d} \right) \mathbf{K}^{-1} \quad [\text{H\&Z, p. 327}]$$

## ► Homography Subgroups: Euclidean Mapping (aka Rigid Motion)

- Euclidean mapping (EM): rotation, translation and their combination

$$\mathbf{H} = \begin{bmatrix} \cos \phi & -\sin \phi & t_x \\ \sin \phi & \cos \phi & t_y \\ 0 & 0 & 1 \end{bmatrix}$$

- eigenvalues  $(1, e^{-i\phi}, e^{i\phi})$   $i = \sqrt{-1}$



**EM = The most general homography preserving**

rotation by  $30^\circ$ , then translation by  $(7, 2)$

1. **areas:**  $\det \mathbf{H} = 1 \Rightarrow$  unit Jacobian

2. **lengths:** Let  $\underline{\mathbf{x}}'_i = \mathbf{H}\underline{\mathbf{x}}_i$  (check we can use  $=$  instead of  $\simeq$ ). Let  $(x_i)_3 = 1$ , Then

$$\|\underline{\mathbf{x}}'_2 - \underline{\mathbf{x}}'_1\| = \|\mathbf{H}\underline{\mathbf{x}}_2 - \mathbf{H}\underline{\mathbf{x}}_1\| = \|\mathbf{H}(\underline{\mathbf{x}}_2 - \underline{\mathbf{x}}_1)\| = \dots = \|\underline{\mathbf{x}}_2 - \underline{\mathbf{x}}_1\|$$

3. **angles** check the dot-product of normalized differences from a point  $(\mathbf{x} - \mathbf{z})^\top (\mathbf{y} - \mathbf{z})$  (Cartesian(!))

- eigenvectors when  $\phi \neq k\pi$ ,  $k = 0, 1, \dots$  (columnwise)

$$\mathbf{e}_1 \simeq \begin{bmatrix} t_x + t_y \cot \frac{\phi}{2} \\ t_y - t_x \cot \frac{\phi}{2} \\ 2 \end{bmatrix}, \quad \mathbf{e}_2 \simeq \begin{bmatrix} i \\ 1 \\ 0 \end{bmatrix}, \quad \mathbf{e}_3 \simeq \begin{bmatrix} -i \\ 1 \\ 0 \end{bmatrix}$$

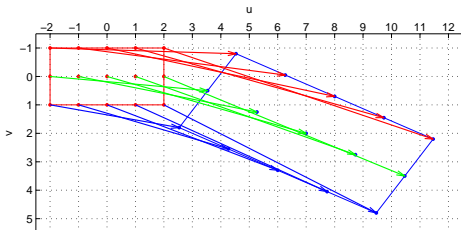
$\mathbf{e}_2, \mathbf{e}_3$  – circular points,  $i$  – imaginary unit

4. **circular points:** points at infinity  $(i, 1, 0)$ ,  $(-i, 1, 0)$  (preserved even by similarity)

- **similarity:** scaled Euclidean mapping (does not preserve lengths, areas)

## ► Homography Subgroups: Affine Mapping

$$\mathbf{H} = \begin{bmatrix} a_{11} & a_{12} & t_x \\ a_{21} & a_{22} & t_y \\ 0 & 0 & 1 \end{bmatrix}$$



### AM = The most general homography preserving

- parallelism
- ratio of areas
- ratio of lengths on parallel lines
- linear combinations of vectors (e.g. midpoints)
- convex hull
- line at infinity  $\underline{n}_\infty$  (not pointwise)

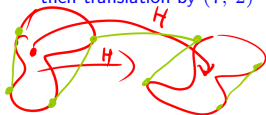
### does not preserve

- lengths
- angles
- areas
- circular points

observe  $\mathbf{H}^T \underline{n}_\infty \simeq \begin{bmatrix} a_{11} & a_{21} & 0 \\ a_{12} & a_{22} & 0 \\ t_x & t_y & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = \underline{n}_\infty \Rightarrow \underline{n}_\infty \simeq \mathbf{H}^{-T} \underline{n}_\infty$

$$\mathbf{H}^T \underline{n}_\infty \simeq \underline{n}_\infty$$

rotation by  $30^\circ$   
then scaling by  $\text{diag}(1, 1.5, 1)$   
then translation by  $(7, 2)$



Euclidean mappings preserve all properties affine mappings preserve, of course



# Elementary Decomposition of a Homography

**Unique decompositions:**  $\mathbf{H} = \mathbf{H}_S \mathbf{H}_A \mathbf{H}_P \quad (= \mathbf{H}'_P \mathbf{H}'_A \mathbf{H}'_S)$

$$\mathbf{H}_S = \begin{bmatrix} s \mathbf{R} & \mathbf{t} \\ \mathbf{0}^\top & 1 \end{bmatrix} \quad \text{similarity (scaled EM)}$$

$$\mathbf{H}_A = \begin{bmatrix} \mathbf{K} & \mathbf{0} \\ \mathbf{0}^\top & 1 \end{bmatrix} \quad \text{special affine}$$

$$\mathbf{H}_P = \begin{bmatrix} \mathbf{I} & \mathbf{0} \\ \mathbf{v}^\top & w \end{bmatrix} \quad \text{special projective}$$

$\mathbf{K}$  – upper triangular matrix with positive diagonal entries

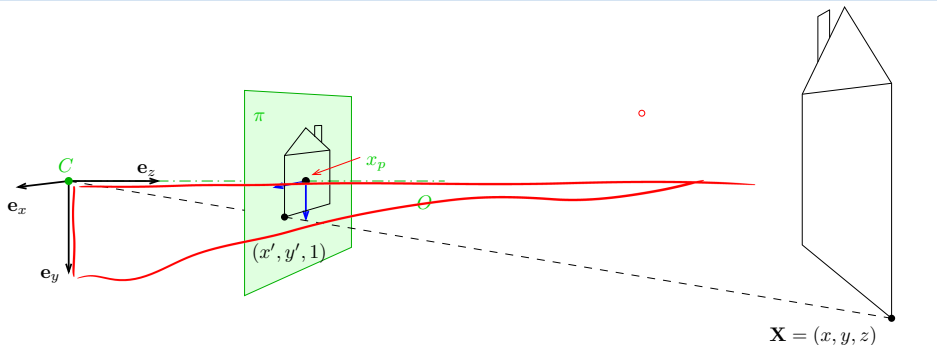
$\mathbf{R}$  – orthogonal,  $\mathbf{R}^\top \mathbf{R} = \mathbf{I}$ ,  $\det \mathbf{R} = 1$

$s, w \in \mathbb{R}$ ,  $s > 0$ ,  $w \neq 0$

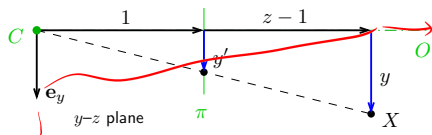
$$\mathbf{H} = \begin{bmatrix} s \mathbf{R} \mathbf{K} + \mathbf{t} \mathbf{v}^\top & w \mathbf{t} \\ \mathbf{v}^\top & w \end{bmatrix}$$

- must use 'thin' QR decomposition, which is unique [Golub & van Loan 2013, Sec. 5.2.6]
- $\mathbf{H}_S$ ,  $\mathbf{H}_A$ ,  $\mathbf{H}_P$  are homography subgroups (in the sense of group theory)  
(eg.  $\mathbf{K} = \mathbf{K}_1 \mathbf{K}_2$ ,  $\mathbf{K}^{-1}$ ,  $\mathbf{I}$  are all upper triangular with unit determinant, associativity holds)

## ► Canonical Perspective Camera (Pinhole Camera, Camera Obscura)



1. in this picture we are looking 'down the street'
2. right-handed canonical coordinate system  $(x, y, z)$  with unit vectors  $e_x, e_y, e_z$
3. origin = center of projection  $C$
4. image plane  $\pi$  at unit distance from  $C$
5. optical axis  $O$  is perpendicular to  $\pi$
6. principal point  $x_p$ : intersection of  $O$  and  $\pi$
7. perspective camera is given by  $C$  and  $\pi$



projected point in the natural image coordinate system:

$$\frac{y'}{1} = y' = \frac{y}{1 + z - 1} = \frac{y}{z}, \quad x' = \frac{x}{z}$$



## ► Natural and Canonical Image Coordinate Systems

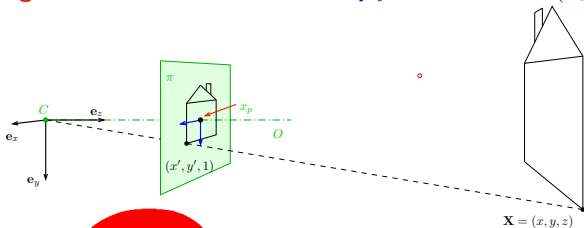
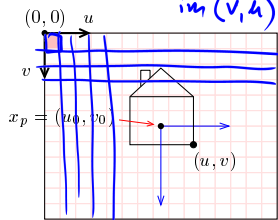
projected point **in canonical camera** ( $z \neq 0$ )

$$(x', y', 1) = \left( \frac{x}{z}, \frac{y}{z}, 1 \right) = \frac{1}{z} (x, y, z) \stackrel{\text{red circle}}{\approx} \underbrace{\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}}_{\mathbf{P}_0} \cdot \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \mathbf{P}_0 \underline{\mathbf{X}}$$

$\mathbf{X} \in \mathbb{R}^3$

projected point **in scanned image**

scale by  $f$  and translate to  $(u_0, v_0)$



$$u = f \frac{x}{z} + u_0$$

$$v = f \frac{y}{z} + v_0$$

$$\frac{1}{z} \begin{bmatrix} f x + z u_0 \\ f y + z v_0 \\ z \end{bmatrix} \approx \underbrace{\begin{bmatrix} f & 0 & u_0 \\ 0 & f & v_0 \\ 0 & 0 & 1 \end{bmatrix}}_{\mathbf{K}} \cdot \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \cdot \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \mathbf{K} \mathbf{P}_0 \underline{\mathbf{X}} = \mathbf{P} \underline{\mathbf{X}}$$

= 15

- 'calibration' matrix  $\mathbf{K}$  transforms canonical  $\mathbf{P}_0$  to standard perspective camera  $\mathbf{P}$

## ► Computing with Perspective Camera Projection Matrix

$$\underline{\mathbf{m}} = \begin{bmatrix} m_1 \\ m_2 \\ m_3 \end{bmatrix} = \underbrace{\begin{bmatrix} f & 0 & u_0 & 0 \\ 0 & f & v_0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}}_{\mathbf{P}} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} \simeq \begin{bmatrix} fx + u_0z \\ fy + v_0z \\ z \end{bmatrix} \simeq \underbrace{\begin{bmatrix} x + \frac{z}{f}u_0 \\ y + \frac{z}{f}v_0 \\ \frac{z}{f} \end{bmatrix}}_{(a)}$$

$$\frac{m_1}{m_3} = \frac{f x}{z} + u_0 = u, \quad \frac{m_2}{m_3} = \frac{f y}{z} + v_0 = v \quad \text{when } m_3 \neq 0$$

$f$  – ‘focal length’ – converts length ratios to pixels,  $[f] = \text{px}$ ,  $f > 0$

$(u_0, v_0)$  – principal point in pixels

### Perspective Camera:

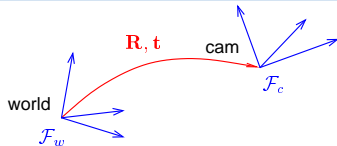
1. dimension reduction since  $\mathbf{P} \in \mathbb{R}^{3,4}$
2. nonlinear unit change  $\mathbf{1} \mapsto \mathbf{1} \cdot z/f$ , see (a)  
for convenience we use  $P_{11} = P_{22} = f$  rather than  $P_{33} = 1/f$  and the  $u_0, v_0$  in relative units
3.  $m_3 = 0$  represents points at infinity in image plane  $\pi$  i.e. points with  $z = 0$

## ► Changing The Outer (World) Reference Frame

A transformation of a point from the world to camera coordinate system:

$$\mathbf{X}_c = \mathbf{R} \mathbf{X}_w + \mathbf{t}$$

$\mathbf{R}$  – camera rotation matrix  
 $\mathbf{t}$  – camera translation vector



world orientation in the camera coordinate frame  $\mathcal{F}_c$   
world origin in the camera coordinate frame  $\mathcal{F}_c$

$$\mathbf{P} \underline{\mathbf{X}}_c = \mathbf{K} \mathbf{P}_0 \begin{bmatrix} \mathbf{X}_c \\ 1 \end{bmatrix} = \mathbf{K} \mathbf{P}_0 \begin{bmatrix} \mathbf{R} \mathbf{X}_w + \mathbf{t} \\ 1 \end{bmatrix} = \mathbf{K} \mathbf{P}_0 \underbrace{\begin{bmatrix} \mathbf{R} & \mathbf{t} \\ \mathbf{0}^\top & 1 \end{bmatrix}}_{\mathbf{T}} \begin{bmatrix} \mathbf{X}_w \\ 1 \end{bmatrix} = \mathbf{K} \underbrace{[\mathbf{R} \mid \mathbf{t}]}_{[\mathbf{R}, \mathbf{t}]} \underline{\mathbf{X}}_w$$

$\mathbf{P}_0$  (a  $3 \times 4$  mtx) selects the first 3 rows of  $\mathbf{T}$  and discards the last row

- $\mathbf{R}$  is rotation,  $\mathbf{R}^\top \mathbf{R} = \mathbf{I}$ ,  $\det \mathbf{R} = +1$
- 6 **extrinsic parameters**: 3 rotation angles (Euler theorem), 3 translation components
- alternative, often used, camera representations

$$\mathbf{P}_0 = [\mathbf{I}_3 \mid \mathbf{0}]$$

$\mathbf{I} \in \mathbb{R}^{3,3}$  identity matrix

$$\mathbf{P} = \mathbf{K} [\mathbf{R} \mid \mathbf{t}] = \mathbf{K} \mathbf{R} [\mathbf{I} \mid -\mathbf{C}]$$

$\mathbf{C}$  – camera position in the world reference frame  $\mathcal{F}_w$   
 $\mathbf{r}_3^\top$  – optical axis in the world reference frame  $\mathcal{F}_w$

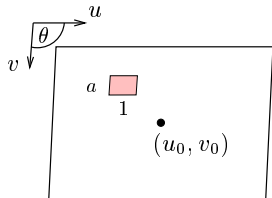
$\mathbf{t} = -\mathbf{R} \mathbf{C}$   
third row of  $\mathbf{R}$ :  $\mathbf{r}_3 = \mathbf{R}^{-1} [0, 0, 1]^\top$

- we can save some conversion and computation by noting that  $\mathbf{K} \mathbf{R} [\mathbf{I} \mid -\mathbf{C}] \underline{\mathbf{X}} = \mathbf{K} \mathbf{R} (\underline{\mathbf{X}} - \mathbf{C})$

## ► Changing the Inner (Image) Reference Frame

The general form of calibration matrix  $\mathbf{K}$  includes

- skew angle  $\theta$  of the digitization raster
- pixel aspect ratio  $a$



$$\mathbf{K} = \begin{bmatrix} f & -f \cot \theta & u_0 \\ 0 & f/(a \sin \theta) & v_0 \\ 0 & 0 & 1 \end{bmatrix}$$

units:  $[f] = \text{px}$ ,  $[u_0] = \text{px}$ ,  $[v_0] = \text{px}$ ,  $[a] = 1$

⊛ H1; 2pt: Verify this  $\mathbf{K}$ . Hints: (1) Map first by skew then by sampling scale then shift by  $u_0, v_0$ ; (2) Skew: express point  $\mathbf{x}$  as  $\mathbf{x} = u' \mathbf{e}_{u'} + v' \mathbf{e}_{v'} = u \mathbf{e}_u + v \mathbf{e}_v$ ,  $\mathbf{e}_u, \mathbf{e}_v$  etc. are unit basis vectors,  $\mathbf{K}$  maps from an orthogonal system to a skewed system  $[w' u', w' v', w']^T = \mathbf{K}[u, v, 1]^T$ ; deadline LD+2 wk

general finite perspective camera has 11 parameters:

- 5 intrinsic parameters:  $f, u_0, v_0, a, \theta$
- 6 extrinsic parameters:  $\mathbf{t}, \mathbf{R}(\alpha, \beta, \gamma)$

finite camera:  $\det \mathbf{K} \neq 0$

$$\underline{\mathbf{m}} \simeq \mathbf{P} \underline{\mathbf{X}}, \quad \mathbf{P} = [\mathbf{Q} \quad \mathbf{q}] = \mathbf{K} [\mathbf{R} \quad \mathbf{t}] = \mathbf{K} \mathbf{R} [\mathbf{I} \quad -\mathbf{C}]$$

a recipe for filling  $\mathbf{P}$

Representation Theorem: The set of projection matrices  $\mathbf{P}$  of finite perspective cameras is isomorphic to the set of homogeneous  $3 \times 4$  matrices with the left  $3 \times 3$  submatrix  $\mathbf{Q}$  non-singular. ↗

Thank You

