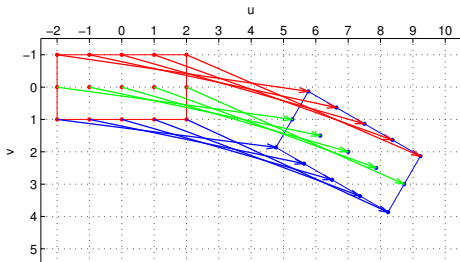


► Homography Subgroups: Euclidean Mapping (aka Rigid Motion)

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



rotation by 30° , then translation by $(7, 2)$

EM = The most general homography preserving

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)

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

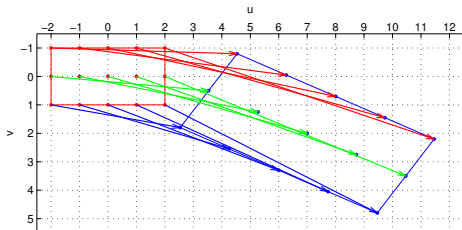
$\mathbf{e}_2, \mathbf{e}_3$ – circular points, i – imaginary unit

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}_∞ (not pointwise)

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

does not preserve

- lengths
- angles
- areas
- circular points

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

Euclidean mappings preserve all properties affine mappings preserve, of course

► Homography Subgroups: General Homography

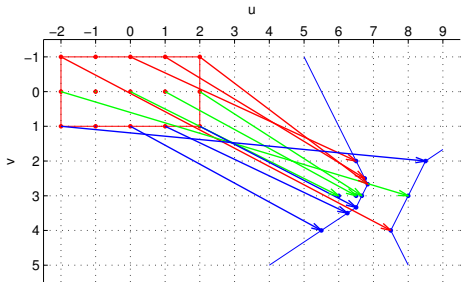
$$\mathbf{H} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix}$$

preserves only

- incidence and concurrency
- collinearity
- cross-ratio on the line $\rightarrow 45$

does not preserve

- lengths
- areas
- parallelism
- ratio of areas
- ratio of lengths
- linear combinations of vectors (midpoints, etc.)
- convex hull
- line at infinity \underline{n}_∞

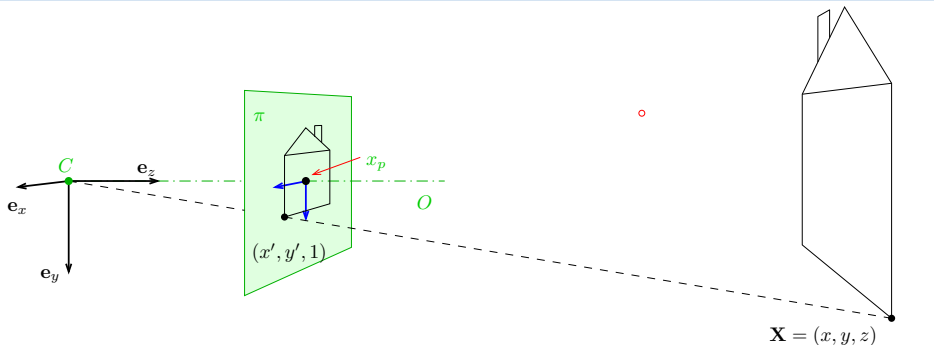


$$\mathbf{H} = \begin{bmatrix} 7 & -0.5 & 6 \\ 3 & 1 & 3 \\ 1 & 0 & 1 \end{bmatrix}$$

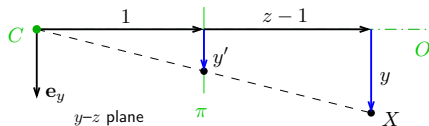
line $\underline{n} = (1, 0, 1)$ is mapped to \underline{n}_∞ : $\mathbf{H}^{-T} \underline{n} \simeq \underline{n}_\infty$

(where in the picture is the line \underline{n} ?)

► 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 π at unit distance from C
5. optical axis O is perpendicular to π
6. principal point x_p : intersection of O and π
7. perspective camera is given by C and π



projected point in the natural image coordinate system:

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

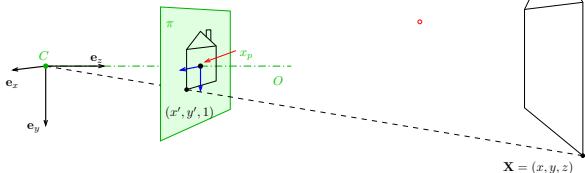
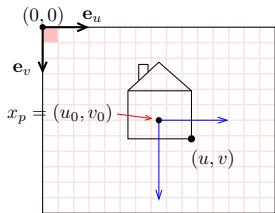
► Natural and Canonical Image Coordinate Systems

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

$$(x', y', 1) = \left(\frac{x}{z}, \frac{y}{z}, 1 \right) = \frac{1}{z}(x, y, z) \simeq \underbrace{\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}}_{\mathbf{P}_0 = [\mathbf{I} \quad \mathbf{0}]} \cdot \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \mathbf{P}_0 \underline{\mathbf{X}}$$

projected point **in scanned image**

scale by f and translate to (u_0, v_0)



$$\begin{aligned} u &= f \frac{x}{z} + u_0 \\ v &= f \frac{y}{z} + v_0 \end{aligned} \quad \frac{1}{z} \begin{bmatrix} f x + z u_0 \\ f y + z v_0 \\ z \end{bmatrix} \simeq \begin{bmatrix} f & 0 & u_0 \\ 0 & f & v_0 \\ 0 & 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \cdot \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \mathbf{K} \mathbf{P}_0 \underline{\mathbf{X}} = \mathbf{P} \underline{\mathbf{X}}$$

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

► Computing with Perspective Camera Projection Matrix

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

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

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 π i.e. points with $z = 0$

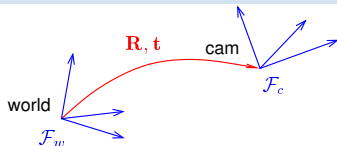
► Changing The Outer (World) Reference Frame

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

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

\mathbf{P}_0 (a 3×4 mtx) discards the last row of \mathbf{T}

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

$$\mathbf{P} = \mathbf{K} \begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix} = \mathbf{K} \mathbf{R} \begin{bmatrix} \mathbf{I} & -\mathbf{C} \end{bmatrix}$$

\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

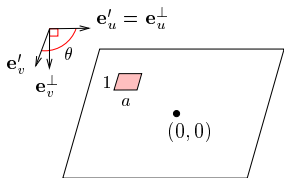
$\mathbf{t} = -\mathbf{R} \mathbf{C}$
third row of \mathbf{R} : $\mathbf{r}_3 = \mathbf{R}^{-1} [0, 0, 1]^\top$

- we can save some conversion and computation by noting that $\mathbf{K} \mathbf{R} \begin{bmatrix} \mathbf{I} & -\mathbf{C} \end{bmatrix} \underline{\mathbf{X}} = \mathbf{K} \mathbf{R} (\underline{\mathbf{X}} - \mathbf{C})$

► Changing the Inner (Image) Reference Frame

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

- skew angle θ of the digitization raster
- pixel aspect ratio a



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

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

⊗ H1; 2pt: Verify this \mathbf{K} . Hints: (1) image projects to orthogonal system F^\perp , then it maps by skew to F' , then by scale f , $a f$ to F'' , then by translation by u_0 , v_0 to F''' ; (2) Skew: express point \mathbf{x} as $\mathbf{x} = u' \mathbf{e}_{u'} + v' \mathbf{e}_{v'} = u^\perp \mathbf{e}_u^\perp + v^\perp \mathbf{e}_v^\perp$, \mathbf{e} : are unit basis vectors, \mathbf{K} maps from F^\perp to F''' as $w''' [u''', v''', 1]^\top = \mathbf{K} [u^\perp, v^\perp, 1]^\top$;

deadline LD+2 wk

general finite perspective camera has 11 parameters:

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

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

$$\underline{\mathbf{m}} \simeq \mathbf{P} \underline{\mathbf{X}}, \quad \mathbf{P} = \begin{bmatrix} \mathbf{Q} & \mathbf{q} \end{bmatrix}; = \mathbf{K} \begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix} = \mathbf{K} \mathbf{R} \begin{bmatrix} \mathbf{I} & -\mathbf{C} \end{bmatrix}$$

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 3×4 matrices with the left 3×3 submatrix \mathbf{Q} non-singular.

► Projection Matrix Decomposition

$$\mathbf{P} = [\mathbf{Q} \quad \mathbf{q}] \longrightarrow \mathbf{K} [\mathbf{R} \quad \mathbf{t}]$$

$\mathbf{Q} \in \mathbb{R}^{3,3}$ full rank (if finite perspective camera; see [H&Z, Sec. 6.3] for cameras at infinity)
 $\mathbf{K} \in \mathbb{R}^{3,3}$ upper triangular with positive diagonal elements
 $\mathbf{R} \in \mathbb{R}^{3,3}$ rotation: $\mathbf{R}^\top \mathbf{R} = \mathbf{I}$ and $\det \mathbf{R} = +1$

1. $[\mathbf{Q} \quad \mathbf{q}] = \mathbf{K} [\mathbf{R} \quad \mathbf{t}] = [\mathbf{KR} \quad \mathbf{Kt}]$ also $\rightarrow 34$
2. RQ decomposition of $\mathbf{Q} = \mathbf{KR}$ using three Givens rotations [H&Z, p. 579]

$$\mathbf{K} = \mathbf{Q} \underbrace{\mathbf{R}_{32} \mathbf{R}_{31} \mathbf{R}_{21}}_{\mathbf{R}^{-1}} \quad \mathbf{Q} \mathbf{R}_{32} = \begin{bmatrix} \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \\ 0 & 0 & \cdot \end{bmatrix}, \quad \mathbf{Q} \mathbf{R}_{32} \mathbf{R}_{31} = \begin{bmatrix} \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \\ 0 & 0 & \cdot \end{bmatrix}, \quad \mathbf{Q} \mathbf{R}_{32} \mathbf{R}_{31} \mathbf{R}_{21} = \begin{bmatrix} \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \\ 0 & 0 & \cdot \end{bmatrix}$$

\mathbf{R}_{ij} zeroes element ij in \mathbf{Q} affecting only columns i and j and the sequence preserves previously zeroed elements, e.g. (see next slide for derivation details)

$$\mathbf{R}_{32} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & c & -s \\ 0 & s & c \end{bmatrix} \text{ gives } \begin{matrix} c^2 + s^2 = 1 \\ 0 = k_{32} = c q_{32} + s q_{33} \end{matrix} \Rightarrow c = \frac{q_{33}}{\sqrt{q_{32}^2 + q_{33}^2}} \quad s = \frac{-q_{32}}{\sqrt{q_{32}^2 + q_{33}^2}}$$

⊛ P1; 1pt: Multiply known matrices \mathbf{K} , \mathbf{R} and then decompose back; discuss numerical errors

- RQ decomposition nonuniqueness: $\mathbf{KR} = \mathbf{KT}^{-1}\mathbf{TR}$, where $\mathbf{T} = \text{diag}(-1, -1, 1)$ is also a rotation, we must correct the result so that the diagonal elements of \mathbf{K} are all positive
‘thin’ RQ decomposition
- care must be taken to avoid overflow, see [Golub & van Loan 2013, sec. 5.2]

RQ Decomposition Step

```
Q = Array [q_{#1,#2} &, {3, 3}];  
R32 = {{1, 0, 0}, {0, c, -s}, {0, s, c}}; R32 // MatrixForm
```

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & c & -s \\ 0 & s & c \end{pmatrix}$$

```
Q1 = Q . R32 ; Q1 // MatrixForm
```

$$\begin{pmatrix} q_{1,1} & c q_{1,2} + s q_{1,3} & -s q_{1,2} + c q_{1,3} \\ q_{2,1} & c q_{2,2} + s q_{2,3} & -s q_{2,2} + c q_{2,3} \\ q_{3,1} & c q_{3,2} + s q_{3,3} & -s q_{3,2} + c q_{3,3} \end{pmatrix}$$

```
s1 = Solve [{Q1[[3]][[2]] = 0, c^2 + s^2 = 1}, {c, s}][[2]]
```

$$\left\{ c \rightarrow \frac{q_{3,3}}{\sqrt{q_{3,2}^2 + q_{3,3}^2}}, s \rightarrow -\frac{q_{3,2}}{\sqrt{q_{3,2}^2 + q_{3,3}^2}} \right\}$$

```
Q1 /. s1 // Simplify // MatrixForm
```

$$\begin{pmatrix} q_{1,1} & \frac{-q_{1,3} q_{3,2} + q_{1,2} q_{3,3}}{\sqrt{q_{3,2}^2 + q_{3,3}^2}} & \frac{q_{1,2} q_{3,2} + q_{1,3} q_{3,3}}{\sqrt{q_{3,2}^2 + q_{3,3}^2}} \\ q_{2,1} & \frac{-q_{2,3} q_{3,2} + q_{2,2} q_{3,3}}{\sqrt{q_{3,2}^2 + q_{3,3}^2}} & \frac{q_{2,2} q_{3,2} + q_{2,3} q_{3,3}}{\sqrt{q_{3,2}^2 + q_{3,3}^2}} \\ q_{3,1} & 0 & \sqrt{q_{3,2}^2 + q_{3,3}^2} \end{pmatrix}$$

► Center of Projection

Observation: finite \mathbf{P} has a non-trivial right null-space

rank 3 but 4 columns

Theorem

Let \mathbf{P} be a camera and let there be $\underline{\mathbf{B}} \neq \mathbf{0}$ s.t. $\mathbf{P} \underline{\mathbf{B}} = \mathbf{0}$. Then $\underline{\mathbf{B}}$ is equivalent to the projection center $\underline{\mathbf{C}}$ (homogeneous, in world coordinate frame).

Proof.

1. Consider spatial line AB (B is given, $A \neq B$). We can write

$$\underline{\mathbf{X}}(\lambda) \simeq \lambda \underline{\mathbf{A}} + (1 - \lambda) \underline{\mathbf{B}}, \quad \lambda \in \mathbb{R}$$

2. it projects to

$$\mathbf{P} \underline{\mathbf{X}}(\lambda) \simeq \lambda \mathbf{P} \underline{\mathbf{A}} + (1 - \lambda) \mathbf{P} \underline{\mathbf{B}} \simeq \mathbf{P} \underline{\mathbf{A}}$$

- the entire line projects to a single point \Rightarrow it must pass through the optical center of \mathbf{P}
- this holds for any choice of $A \neq B \Rightarrow$ the only common point of the lines is the C , i.e. $\underline{\mathbf{B}} \simeq \underline{\mathbf{C}}$

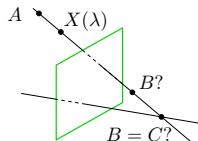
□

Hence

$$\mathbf{0} = \mathbf{P} \underline{\mathbf{C}} = [\mathbf{Q} \quad \mathbf{q}] \begin{bmatrix} \underline{\mathbf{C}} \\ 1 \end{bmatrix} = \mathbf{Q} \underline{\mathbf{C}} + \mathbf{q} \Rightarrow \underline{\mathbf{C}} = -\mathbf{Q}^{-1} \mathbf{q}$$

$\underline{\mathbf{C}} = (c_j)$, where $c_j = (-1)^j \det \mathbf{P}^{(j)}$, in which $\mathbf{P}^{(j)}$ is \mathbf{P} with column j dropped

Matlab: `C_homo = null(P)`; or `C = -Q\q`;



Thank You

