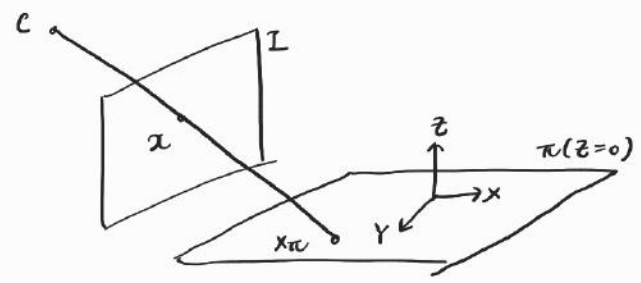


• Action of projective camera on planes

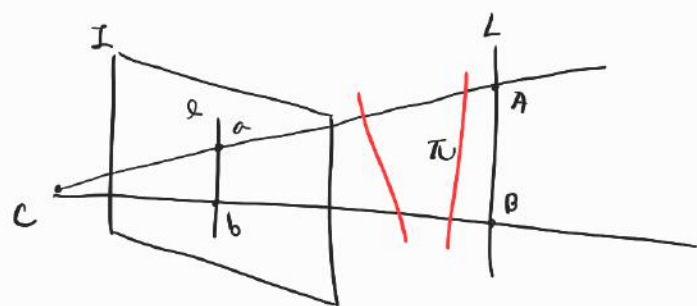


$$x = PX = [p_1 \ p_2 \ p_3 \ p_4] \begin{pmatrix} x \\ y \\ 0 \\ 1 \end{pmatrix} = [p_1 \ p_2 \ p_4] \begin{pmatrix} x \\ y \\ 1 \end{pmatrix}$$

Homography matrix (full rank 3)
 $P^2(\text{xy-plane}) \mapsto P^2(\text{image})$

H calculated using 4 point correspondences \therefore mapping b/w $x_\pi = \begin{pmatrix} x \\ y \\ 1 \end{pmatrix}$ and x is planar homography.
 (plane-to-plane)

• Action of projective camera on lines



• Forward projection

3D-space line (L) \mapsto image line (l)

line l = intersection of π and image plane

Since $L = X(u) = A + uB$,

$$l = x(u) = PX(u) = PA + P u B = a + u b$$

\Rightarrow l is line joining a and b

• Backward projection

Impossible to recover L directly from l

\hookrightarrow Every line in π (e.g. red lines) projects to l

Set of points mapping to line l is the plane $\pi = P^T l$

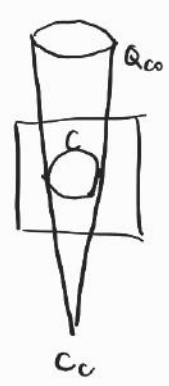
\therefore if point x lies on l, $x^T l = 0$

$$\text{Since } x = PX, \quad x^T l = (PX)^T l = x^T \underbrace{P^T l}_{=\pi} = 0$$

$\therefore x^T \pi = 0 \rightarrow$ point x in 3-space lies on π

• Action of Projective camera on conics

• Back projection of conics



$$Q_{co} = P^T C P \quad \text{where } Q_{co} \in R^{4 \times 4}, C \in R^{3 \times 3}$$

\hookrightarrow Conic C back-projects to degenerate quadric (cone) Q_{co}

$\therefore Q_{co}$ isn't full rank

$$\therefore x^T C x = 0 \quad \text{and } x = PX$$

$$\Rightarrow (PX)^T C (PX) = x^T (P^T C P) x = 0 \quad \therefore Q_{co} = P^T C P \quad \text{Eq(1)}$$

\hookrightarrow Camera center is null-vector of quadric i.e. $Q_{co} C_c = 0 \quad \text{Eq(2)}$

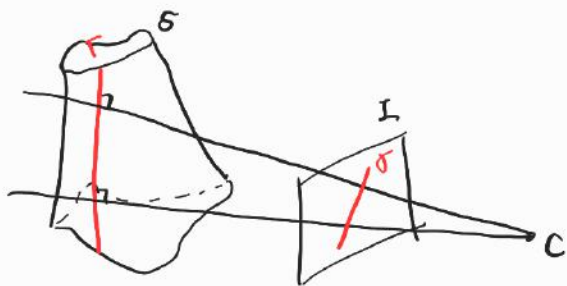
With Eq(1) and Eq(2), $Q_{C0}C = (P^T C P)C = \underbrace{P^T C (P C)}_{=0} = 0$

e.g. if $P = K[I|0]$ (canonical frame : camera = world frame \therefore Extrinsic = I)

$$Q_{C0} = P^T C P = \begin{bmatrix} K^T \\ 0^T \end{bmatrix} C [K|0] = \begin{bmatrix} K^T C K & 0 \\ 0^T & 0 \end{bmatrix}$$

$$\therefore \text{rank}(Q_{C0}) = 3$$

• Images of smooth surfaces



Central projection
(projective camera)

Image outline of smooth surface S

- forward projection of surface points (tangent to rays)
- backward projection of tangent planes to surface

• Contour generator Γ

set of points x on S that are tangent to rays
depends on position of C and S

• apparent contour (outline/profile) δ

set of points x that are image of X
depends on position of I

• Action of projective camera on quadrics

• Forward projection of quadric



$C^* = P Q^* P^T$ i.e. defined by tangent of $\text{lines}(C^*)$ and $\text{planes}(Q^*)$

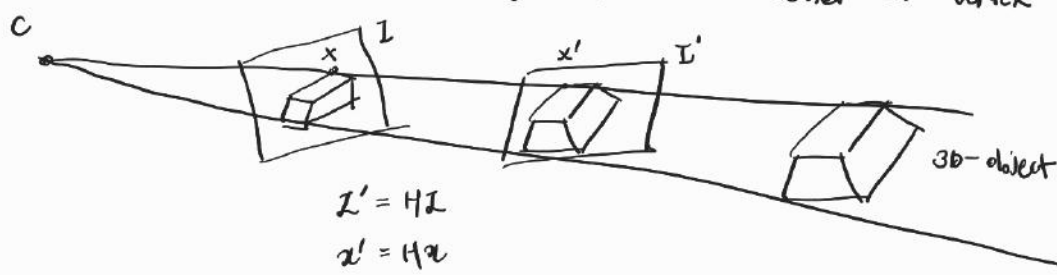
$$\therefore l^T C^* l = \pi^T Q^* \pi = 0 \quad \text{and} \quad l = P \pi \Rightarrow \pi = P^T l$$

$$\pi^T Q^* \pi = \underbrace{l^T P Q^* P^T l}_{C^* = P Q^* P^T} = l^T C^* l = 0$$

$$C^* = P Q^* P^T$$

• Importance of camera center

set of rays : defined by 3D object and camera center \rightarrow image = rays intersecting p^2 plane
↳ referred to cone of rays w/ camera center as vertex



* Projection affected by $R, C, f, I \dots$

Images (I, I') mapped from same camera center (c) can be defined as homography

\Rightarrow Projective equivalent

Let $P = KR[I|-\tilde{C}]$, $P' = K'R'[I|-\tilde{C}]$ i.e. camera center is equal

$$\therefore P' = (K'R')(KR)^{-1}P$$

$$\text{Similarly, } x' = P'X = (K'R')(KR)^{-1}PX = \underline{(K'R')(KR)^{-1}}x$$

homography H ($P^2 \mapsto P^2$)

(i) Translation of image plane (sf)

\hookrightarrow simple magnification effect

$$x = PX = K[I|0]X$$

$$x' = P'X = \underline{K'[I|0]}X = K'(K^{-1}K)[I|0]X = K'K^{-1}(K[I|0]X) = K'K^{-1}x$$

only f differs

$$\therefore x' = Hx \text{ where } H = K'K^{-1}$$

$$H = K'K^{-1} = \begin{bmatrix} rI & (1-r)\tilde{x}_0 \\ 0^T & 1 \end{bmatrix} \text{ where } \tilde{x}_0 = (\tilde{x}_0, \tilde{y}_0)^T$$

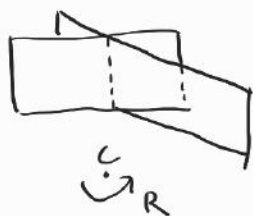
$r = f'/f$ is magnification factor

$$\therefore K^{-1} = \begin{bmatrix} rI & (1-r)\tilde{x}_0 \\ 0^T & 1 \end{bmatrix} K = \begin{bmatrix} rI & (1-r)\tilde{x}_0 \\ 0^T & 1 \end{bmatrix} \begin{bmatrix} A & \tilde{x}_0 \\ 0^T & 1 \end{bmatrix} \text{ where } A = \text{diag}(f, f, 1)$$

$$= \begin{bmatrix} rA & \tilde{x}_0 \\ 0^T & 1 \end{bmatrix} = K \begin{bmatrix} rI & \\ & I \end{bmatrix}$$

\Rightarrow effect of zooming by factor $r = f/f'$

(ii) Camera rotation



Rotation at center C w/o change in internal parameters

$$x = PX = K[I|0]X \text{ e.g. canonical case}$$

$$x' = P'X = K[R|0]X = KR(K^{-1}K)I[R|0]X = KRK^{-1}x$$

$$\therefore H = KRK^{-1} \text{ called conjugate rotation}$$

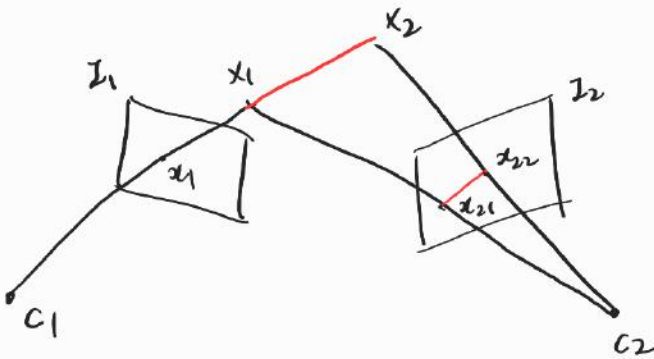
$\hookrightarrow H, R$ shares equal eigenvalues up to scale $\{u, ue^{i\theta}, ue^{-i\theta}\}$

steps: take 4 point correspondences of two image views \rightarrow calculate H

→ Calculate eigenvalue of H → Derive θ

(iii) Moving camera center (motion parallax)

fixed camera center (case (i), (ii)) : No info. of 3D space structure
only $x' = Hx$



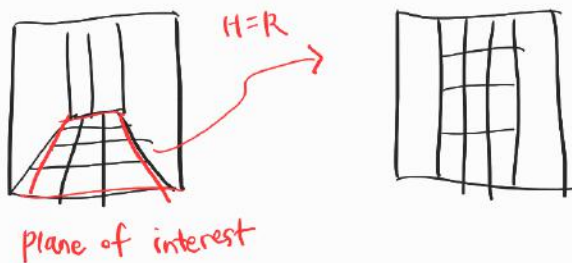
$\Delta C (C_1 \rightarrow C_2)$

Image points now also depends on 3D point
($x_1 \rightarrow x_2 \mapsto x_{21} \rightarrow x_{22}$ in I_2)

• Applications and Examples (fixed camera center)

Ex(1) Synthetic views

warping planar homography from plane of interest



• Algorithm

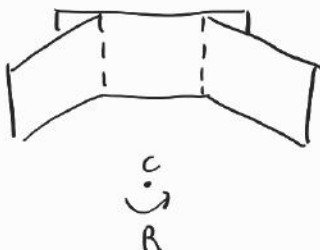
1) Compute H from quadrilateral (source) to rectangle (target) w/ known aspect ratio
↳ usually take 4 corners as point correspondences ($x' = Hx$)

2) projectively warp source image

↳ lookup table $x = H^{-1}x'$ so that all pixels of target image contains a value

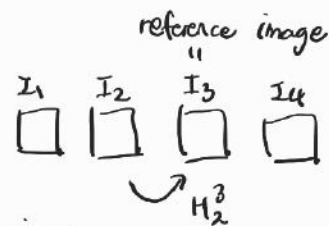
Ex(2) planar panoramic Mosaicing

set of images that camera rotates at center → panorama image stitching



Algorithm

- 1) choose a reference image
- 2) calculate H that maps one image to reference image
- 3) warp non-reference images using H , augment non-overlapping regions of reference image



* If no point correspondences w/ ref. image, use property

eg. $H_1^3 = H_1^2 \cdot H_2^3 \quad \because$ linear mapping

Usage of Calibration $K_{3 \times 3}$

(i) Direction of rays

In E^3 space, points on the ray : $\tilde{x} = \lambda d$ (inhomogeneous form)

Points \tilde{x} map into $x = K[I|0] \underbrace{(\lambda d^T, 1)^T}_x = Kd$

$\therefore d = K^{-1}x \rightarrow$ direction of ray obtained from image point (x unit vector)

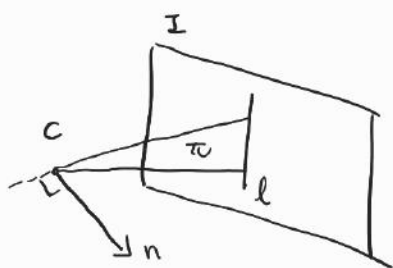
(ii) Angle b/w two rays

$$\cos \theta = \frac{d_1^T d_2}{\|d_1\| \|d_2\|} = \frac{(K^{-1}x_1)^T (K^{-1}x_2)}{\|K^{-1}x_1\| \|K^{-1}x_2\|} = \frac{x_1^T (K^{-T} K^{-1}) x_2}{\sqrt{x_1^T (K^{-T} K^{-1}) x_1} \sqrt{x_2^T (K^{-T} K^{-1}) x_2}}$$

\therefore Calibrated camera work as directional sensor / 2d protractor

\rightarrow find angle b/w any two image points

(iii) Normal of back-projected plane



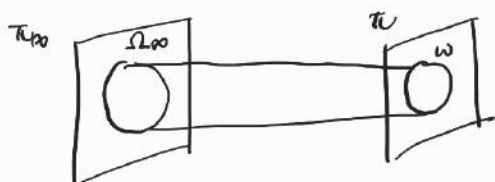
$$n = K^T l$$

\because Points x on l back-projects to $d = K^{-1}x$

$$d \perp n \Rightarrow d^T n = (K^{-1}x)^T n = x^T (K^{-T} n) = 0$$

$$\text{Since } x^T l = 0, \quad l = K^{-T} n \quad \therefore n = K^T l$$

(iv) The image of absolute conic



ω : image of absolute conic (IAC)

Not actually visible at real plane (\because all imaginary points)

points on π_∞ : $x_\infty = (d^T, 0)^T$ w/ general camera $p = KR[I - \tilde{C}]$,

$$x = Px_\infty = KR[I - \tilde{C}] \begin{pmatrix} d \\ 0 \end{pmatrix} = \overbrace{KRd}^{p^2 \rightarrow p^2 \text{ homography } H} \quad \therefore \text{independent of } c/t$$

$$\omega(IAC) = (KK^T)^{-1} = K^{-T}K^{-1}$$

$$\therefore \text{for } x' = Hx, \quad C' = H^{-T}CH^{-1}$$

$$\text{Since } \Omega_\infty \text{ is } C = I \text{ at } \pi_\infty, \quad \omega = H^{-T}CH^{-1} = (KR)^{-T}I(KR)^{-1} = K^{-T}R R^{-T}K^{-1} = (KK^T)^{-1}$$

↳ from equation (ii)

$$\cos \theta = \frac{x_1^T \omega x_2}{\sqrt{x_1^T \omega x_1} \sqrt{x_2^T \omega x_2}} \quad \rightarrow \quad \text{unchanged under projective transformation}$$

$$x_1^T \omega x_2 \xrightarrow{H} x_1'^T \omega' x_2'$$

$$= (x_1^T H^T)(H^{-T} \omega H^{-1})(H x_2) = x_1^T \omega x_2$$

$$\downarrow$$

$$\text{if } x_1 \perp x_2, \quad x_1^T \omega x_2 = 0$$

* Dual image of absolute conic (DIAC)

$$C^* = C^{-1} \text{ when } C \text{ is full rank (non-degenerate)}$$

Since $\omega = K^{-T}K^{-1}$, since K is full rank, ω is always full rank

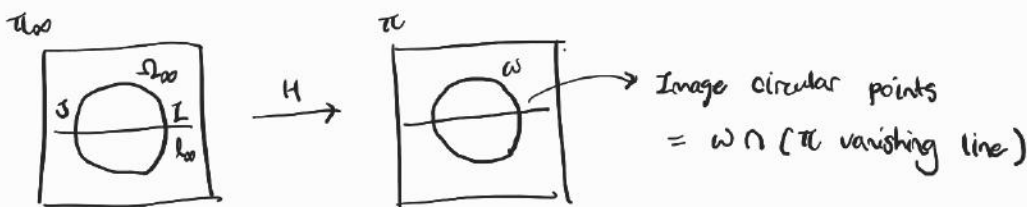
$$\therefore \omega^*(DIAC) = \omega^{-1} = KK^T$$

↳ This is dual (line) conic, although not visible in real plane

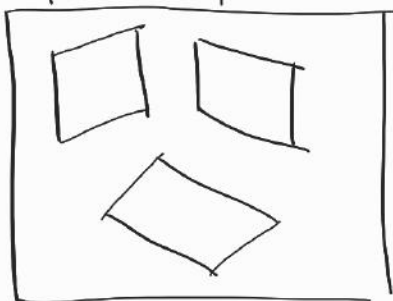
$$\text{Also, } \omega^* = \text{image of } \Omega_\infty^* (\text{dual absolute conic}) = p \Omega_\infty^* p^T$$

* Given ω^* , K can be uniquely derived using Cholesky decomposition

$$\text{cholesky}(\omega^*) = UL = KK^T$$



• Example : A Simple Calibration Device



If three planes aren't parallel, able to find K (calibration)

• Calibration algorithm

1) For each square, compute H that maps 4 corners to image points

* Circular points are invariant to similarity transformation

∴ Absolute scale of rectangle / corner is x important

2) Compute imaged circular points using $H = [h_1, h_2, h_3]$

$$H \cdot \text{Circular points at } \pi_{\infty} = H(1, \pm i, 0)^T = h_1 \pm i h_2$$

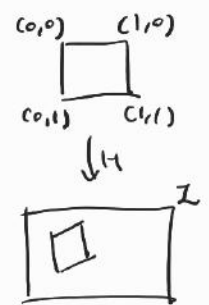
3) Since imaged circular points lies on ZAC:

$$(h_1 \pm i h_2)^T \omega (h_1 \pm i h_2) = 0 \quad \left[\begin{array}{l} h_1^T \omega h_2 = 0 \\ h_1^T \omega h_1 = h_2^T \omega h_2 \end{array} \right] \quad \begin{array}{l} \text{1 pair of circular points} \\ \text{give 2 constraints} \end{array}$$

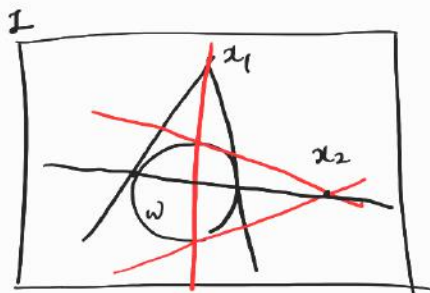
linear r.t. ω

ω has 6 unknowns (5 dof), requires three unique homography / square

4) Compute K using Cholesky factorization of $\omega = (KK^T)^{-1}$

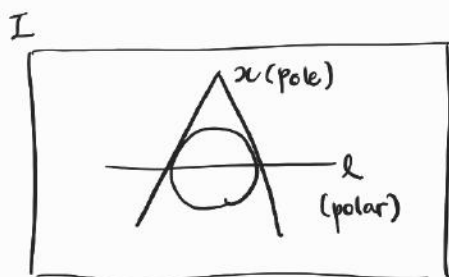


• Orthogonality and ω



x_1, x_2 back-project to orthogonal rays if they are conjugate r.t. $\omega \rightarrow x_1^T \omega x_2 = 0$

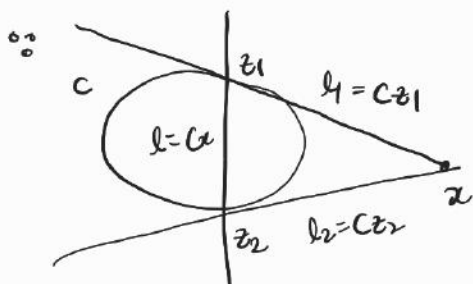
↓ x_2 lies on l



Back-projection of x (ray) and l (plane) is orthogonal if they are pole-polar $\rightarrow l = \omega x$

• Pole-polar relationship

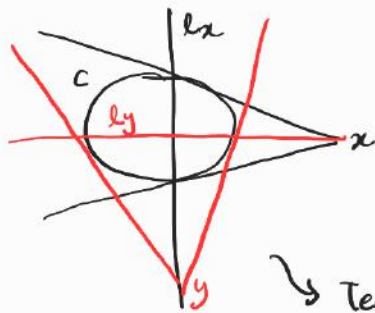
if $l = Cx$, $\left[\begin{array}{l} x^T C x = 0 : l \text{ is tangent line} \\ x^T C x \neq 0 : l \text{ is polar line (pass two tangent points)} \end{array} \right.$



$$x = l_1 \times l_2 = (Cz_1) \times (Cz_2) = \det(C) \underbrace{(C^{-1})^T}_{\substack{\text{C is symmetric} \\ = C^{-1}}} (z_1 \times z_2) = \det(C) C^{-1} l = k C^{-1} l$$

$$l = Cx$$

- Conjugate points



As y lies on $lx = Cx$, $y^T lx = y^T Cx = 0$

$\therefore y^T Cx = 0$ means x, y are conjugate w.r.t. C

* $y^T Cx = x^T Cy = 0$ (symmetric)

Terminology: $(x, lx), (y, ly)$ are (pole, polar) pair
 x, y are conjugate

* Dual conjugacy for lines: lines l, m are conjugate if $l^T C^* m = 0$

- Vanishing points

Intersection of image plane w/ ray parallel to world line passing camera center depends only on ray direction \rightarrow Parallel lines share single vanishing point

Join of any 2 vanishing points \rightarrow vanishing line ($l = v_1 x v_2$)

- Algebraic interpretation



$$x(\lambda) = A + \lambda D \quad \text{where } D = (d^T, 0)^T$$

$$x(\lambda) = P x(\lambda) = K [I | 0] x(\lambda) = K [I | 0] A + \lambda K [I | 0] D = a + \lambda k_d$$

vanishing point is at:

$$v = \lim_{\lambda \rightarrow \infty} x(\lambda) = \lim_{\lambda \rightarrow \infty} a + \lambda k_d = \lim_{\lambda \rightarrow \infty} \frac{a}{\lambda} + k_d = k_d$$

$\Rightarrow v$ depends on direction d , not position A

- Geometric interpretation

Image of intersection b/w π_{∞} and set of parallel lines w/ direction d

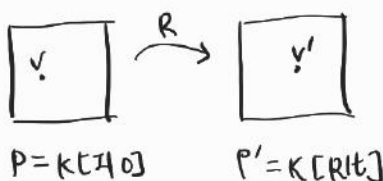
$$v = P x_{\infty} = K [I | 0] \begin{pmatrix} d \\ 0 \end{pmatrix} = k_d$$

* parallel image lines might not be parallel in 3-space

e.g. any intersecting lines at principal plane ($z=0$) imaged as parallel lines

\therefore intersecting point is x_{∞}

e.g. rotation estimation from vanishing points



$$v = P x_{\infty} = K [I | 0] (d^T, 0)^T = k_d \quad \therefore d = K^{-1} v$$

$$v' = P' x_{\infty} = K [R | t] (d^T, 0)^T = K R d = k_{d'} \quad \therefore d' = K^{-1} v'$$

Since K, V, V' is known, d, d' can be calculated \rightarrow Derive R from $d' = Rd$

* $d = \begin{pmatrix} dx \\ dy \\ 1 \end{pmatrix} = K^{-1}V$ gives 2 constraints & R has 3 dof

\Rightarrow it requires at least 2 (V, V') pairs to find R

eg. angle b/w 2 scene lines

$$\cos \theta = \frac{V_1^T W V_2}{\sqrt{V_1^T W V_1} \sqrt{V_2^T W V_2}}$$

• chicken-and-egg problem

Set of imaged parallel scene lines $\xrightarrow{\text{intersection of lines}}$ vanishing points
 $\xleftarrow{\text{lines passing } v}$

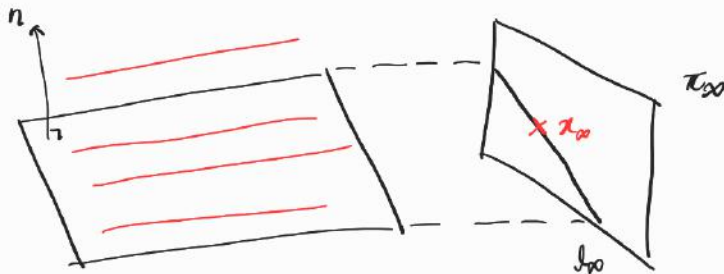
\Rightarrow Both are unknown!
 & detected lines inaccuracy due to noise

• Vanishing Lines

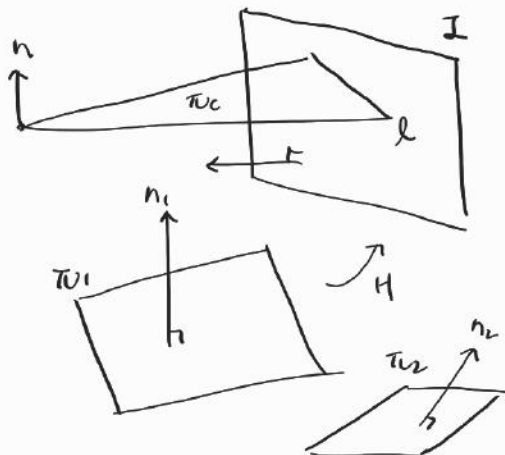
\hookrightarrow Image of line of infinity, which is the intersection of plane of infinity and set of parallel planes in 3-space.

\hookrightarrow OR, intersection of image plane and plane parallel to scene plane that contains camera center

\therefore Dependent on camera center & plane orientation



• Usages (w/ known K)



π (plane w/ vanishing line l)

$$: n = K^T l$$

π_1 and I are related as plane-to-plane homography
 $: \text{metrically rectify from } n_1 \text{ to fronto-parallel scene}$

Determine angle b/w 2 scene planes

$$: \text{since } n_1 = K^T l_1, n_2 = K^T l_2.$$

$$\cos \theta = \frac{l_1^T W^* l_2}{\sqrt{l_1^T W^* l_1} \sqrt{l_2^T W^* l_2}}$$

- Computing vanishing lines

1) Find V_1, V_2 from 2 set of parallel lines

2) $L = V_1 \times V_2$

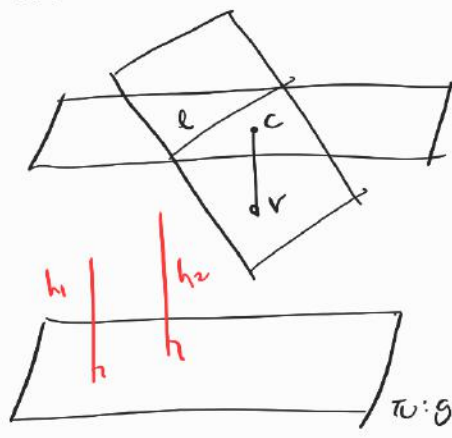
- Orthogonality relationships

if V_1, V_2 have perpendicular lines, $V_1^T W V_2 = 0$

if L, V have perpendicular plane/line, $L = W V$ or $V = W^* L$ ($\because W^{-1} = W^*$)

if L_1, L_2 have perpendicular planes, $L_1^T W^* L_2 = 0$

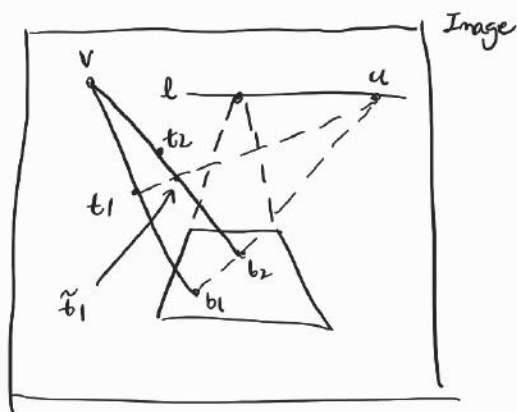
- Affine 3D Measurements



Given vanishing point V and vanishing line L ,
Compute ratio of vertical line segments h_2/h_1

$$h_1 \rightarrow \overline{b_1 t_1}$$

$$h_2 \rightarrow \overline{b_2 t_2}$$

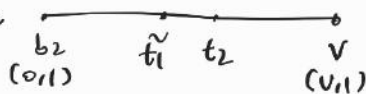


- steps

1) $u = (b_1 \times b_2) \times L$

2) $\tilde{t}_1 = (t_1 \times u) \times (v \times b_2)$

3) Define P^1 space



4) Compute $H_{2 \times 2}$ using $(0,1) \rightarrow (0,1)$ and $(v,1) \rightarrow (1,0)$ mapping
(origin: $b_2 \rightarrow b_2$) ($v \rightarrow x_{\infty}$)

$$\therefore H_{2 \times 2} = \begin{bmatrix} 1 & 0 \\ 1 & -v \end{bmatrix}$$

$$5) \frac{h_2}{h_1} = \frac{H_{2 \times 2} (b_2, 1)^T}{H_{2 \times 2} (\tilde{t}_1, 1)^T}$$

- Determining K

- Linear constraints

$$V_1^T W V_2 = 0 : V_1 \perp V_2$$

$$L \times (W V) = 0 : L \perp V$$

$$h_1^T W h_2 = 0, h_1^T W h_1 = h_2^T W h_2 : \text{known } H$$

$$W_{12} = W_{21} = 0 : \text{Zero skew}$$

+

$$W_{11} = W_{22} : \text{Square pixels}$$

• Steps

1) Define $w \in \mathbb{R}^{1 \times 6}$ with 5 dof

2) Stack above linear constraints $a^T w = 0 \rightarrow A^T w = 0$ where $A \in \mathbb{R}^{n \times 6}$

3) Solve w , find K using cholesky decomposition : $SVD(w) = (K K^T)^{-1}$