

- 1 BILINEAR EXERCISE
- 2 CONVERT MAX SPACE TO PPI
- 3 SAMPLING & QUANTIZATION
- 4 SUPERSAMPLING
- 5 QUANTIZATION ISSUES AND SOLUTIONS
- 6 BILINEAR INTERPOLATION
- 7 GEOMETRIC TRANSFORMATIONS
- 8 HOMOGENEOUS COORDINATES
TRANSFORMATIONS (AFFINE)
- 9 FORWARD MAPPING
- 10 INVERSE MAPPING
- 11 3-SHEARS ROTATION
- 12 CHANGE INTENSITY RESOLUTION
- 13 TEMPORAL INTENSITY SUPER-SAMPLING
- 14 INTENSITY TRANSFORMATIONS
- 15 HISTOGRAM EQUALIZATION
- 16 HISTOGRAM MATCHING
- 17 LINEAR SPATIAL FILTERING
- 18 MEDIAN FILTER
- 19 LAPLACIAN FILTER (smoothing)
- 20 PREWITT FILTER (Box smoothing)
- 21 SOBEL FILTER
- 22 BASIC BINARIZATION LINES
- 23 IMAGE SEGMENTATION BASICS
- 24 SIGNAL SMOOTHING & AVG. SMOOTHING
- 25 EDGE DETECTION (BASIC CONCEPT)
- 26 EDGE DETECTION WITH FIRST ORDER DER.
- 27 EDGE PROPERTIES
- 28 CANNY EDGE DETECTION
- 29 NON-MAXIMA SUPPRESSION
- 30 HYSTERESIS TRESHOLDING
- 31 CONNECTIVITY ANALYSIS
- 32 HOUGH TRANSFORM HORIZONTAL

- (33) HOUGH TRANSFORM ON LINES ON ANY CHOSEN ANGLE α
- (34) LINE HOUGH TRANSFORM (COMPLETE)
- (35) HOUGH CIRCLES
- (36) HARRIS-STEVENS CORNER DETECTOR
- (37) HARRIS-STEVENS WITH EIGENVAL
- (38) COLOR XYZ
- (39) COLOR xyY
- (40) RGB \leftrightarrow XYZ
- (41) RGB \leftrightarrow CMY
- (42) CMY \leftrightarrow CMYK
- (43) RGB \leftrightarrow HSI
- (44) HSI \leftrightarrow RGB
- (45) PSEUDO - COLORS
- (46) CROSS - CORRELATION vs CONVOLUTION
- (47) STRUCTURAL PATTERN MATCHING
- (48) CONIC SECTION
- (49) CONIC TYPES BASED ON DETERMINANT
- (50) CONIC IN P^2 HOMOGENEOUS
- (51) CONICS & DEGREES OF FREEDOM
- (52) DEGENERATE CONIC
- (53) DEGENERATE CONICS IDENTIFICATION
- (54) PROOF OF $C = l m^T + m l^T$ IS THE DEGENERATE CONIC COMPOSED OF LINES l and m
- (55) TANGENT LINES TO CONIC (PROOF)
- (56) ADJOINT OF A CONIC
- (57) Proof that $l^T C^* l = 0$ CAPTURES ALL LINES IN C^*
- (58) DESCRIBE P^n
- (59) PROJECTIVE TRANSFORMATIONS IN 2D

(60) PROVE 3 COLLINEAR POINTS STAY COLLINEAR

UNDER A PROJECTIVE TRANSFORMATION

(61) FIND MAPPING WITH $H: 4 \text{ points} \mapsto 4 \text{ points}$

(62) TRANSFORM A POINT x TO POINT x'

(63) TRANSFORM A LINE l TO l'

(64) CONIC C TO CONIC C'

(65) DUAL CONIC TRANSFORM PROOF

(66) SPECIAL CASE: ISOMETRIES

(67) EUCLIDEAN TRANSFORMATIONS (DISPLACEMENT)

(68) SIMILARITIES (SHAPE-PRESERVING)

(69) AFFINE TRANSFORMATIONS

(70) COMPACT REPRESENTATION OF AFFINE TRANS

(71) DECOMPOSE LINEAR PART OF AFFINE TRANS
IN ROTATION AND SCALINGS

(72) AFFINE MAPPING, SOHW THAT INFINITY
LINE MAPPED TO INFINITY LINE

(73) PROJECTIVE TRANSFORMATION

(74) PROJECTIVE TRANSFORMATION LINES AT ∞

(75) TRANS PROPERTIES (WHAT TO CHOOSE?)

(76) CAMERA BASICS

(77) PINHOLE CAMERA BASIC

(78) PINHOLE CAMERA MAPPING

(79) CENTRAL PROJECTION FOR PINHOLE

(80) NON-CENTRAL ORIENTATION PROJECTION FOR PINHOLE

(81) GET WORLD COORD OF POINT FROM
CAMERA COORD

(82) GO FROM WORLD TO PIXEL

(83) PINHOLE CAMERA WITH NON-SQUARE
PIXELS

- (84) SKewness
- (85) GIVEN ONLY P : CAMERA PROPERTIES
- (86) PROVE C IS CAMERA CENTER
AND IS NULL-SPACE OF P
- (87) COMPUTE CAMERA CENTER C SINGULAR
- (88) COMPUTE CAMERA CENTER C NON-SINGULAR
- (89) RETRIEVE K AND R FROM P
- (100) FIND F IF WORLD COORDS ARE ALIGNED TO FIRST CAMERA

- (101) $[e']_x (K' R' K'^{-1}) \Rightarrow (K')^{-T} R' K^T [e]_x$
- (102) GENERAL CASE OF FINDING F
- (103) PROPERTIES OF F ($x'^T F x = 0$)
- (104) PROPERTIES OF F $\begin{cases} F \text{ is unique } 3 \times 3 \text{ Homo} \\ \text{MATRIX WITH RANK = 2} \\ \text{WITH PROPERTY } x'^T F x = 0 \end{cases}$

- (105) OTHER PROPERTIES OF F :
- (106) $F e = 0 = e'^T F$ PROOF
- (107) DECOMPOSE F MATRIX
- (108) PROVE e, e' EPIPOLES LIE ON F_s
THE STEINER CONIC
- (109) FIND TANGENT TO F_s STEINER CONIC
ON e and e' and X_2 (intersection of them)
- (110) PROOF THAT X_2 is POLE OF ℓ_2
 ℓ_2 IS POLAR OF X_2
- (111) GEOMETRIC CONSTRUCTION OF
EPIPOLAR LINE $\ell' = F_x$
- (112) PROJECTIVE INVARIANCE (\hat{F})
OF F FUNDAMENTAL MATRIX PROOF

(113) $(P, P') = (PH, P'H)$
F IS INDEPENDENT FROM WORLD COORD.

(114) FIND MATRIX H THAT
GIVES CANONICAL FORM
OF (P, P')

(115) ASSUMING $P' = [M' | P_4']$ AND WE ARE
IN CANONICAL $\Rightarrow F = [P_4' | M']$

(116) GIVEN F, FIND (P, P') IN
CANONICAL FORM

(117) IF F IS FUNDAMENTAL MATRIX
OF (P, P') and (\tilde{P}, \tilde{P}') THEN
 $\tilde{P} = PH$ AND $\tilde{P}' = P'H$ for some H

(118) GIVEN F AND A SKEW-SYMMETRIC $S \in \mathbb{R}^{3x3}$
THEN F IS THE FUNDAMENTAL OF
 (P, P') FOR $P = [I | 0]$, $P' = [SF | e']$

(119) FUNDAMENTAL MATRIX
PROPERTIES RECAP

(120) ESTIMATE FUNDAMENTAL F

(121) 8-POINT ALGORITHM (START)

(122) 8-POINT ALGORITHM (FIND F)
SINGULAR

(123) LEAST SQUARE SOLUTION
TO FIND F IN 8-POINTS IF $\text{rank} > 8$
(OVERDEFINED)

(124) FIND \tilde{F} CLOSEST TO F IN 8 POINT
IF $\text{rank} = 3$

(125) 7-POINT ALGORITHM

(126) 3D RECONSTRUCTION
USING F AND CANONICAL CAMERA

(127) 3D RECONSTRUCTIONS ($P \leftrightarrow \tilde{P}$ USING H)

(128) CORRECT PROJECTION

① BILINEAR EXERCISE

SCALE FACTOR OF $C=0.25$

BILINEAR INTERPOLATION

For $i-j$ th pixel in target

$$x = (i+0.5) / 0.25$$

$$y = (j+0.5) / 0.25$$

$$k = \text{round}(x) - 1$$

$$l = \text{round}(y) - 1$$

$$u = x - (k+0.5)$$

$$v = y - (l+0.5)$$

$$\partial_{ij}^2 = \text{round} \left((1-u)(1-v) \partial_{k,l}^1 + (1-u)v \partial_{k+1,l}^1 + v(1-v) \partial_{k,l+1}^1 + vu \partial_{k+1,l+1}^1 \right)$$

$$= (1-u)((1-v)\partial_{k,l}^1 + (v)\partial_{k,l+1}^1) +$$

$$(u)((1-v)\partial_{k+1,l}^1 + (v)\partial_{k+1,l+1}^1)$$

WHERE FOR U

$$U = x - (K + 0.5) \quad K = \text{round}(x) - 1$$

$$x = \frac{(i + 0.5)}{0.25}$$

$$\Rightarrow U = \frac{(i + 0.5)}{0.25} - (K + 0.5)$$

$$= \frac{(i + 0.5)}{0.25} - (\text{round}(x) - 1 + 0.5)$$

$$= \frac{(i + 0.5)}{0.25} - \left(\text{round} \left(\frac{i + 0.5}{0.25} \right) - 1 + 0.5 \right)$$

$$= \left(i + \frac{1}{2} \right) \cdot 4 - \text{round} \left(\left(i + \frac{1}{2} \right) 4 \right) + \frac{1}{2}$$

$$= 4i + 2 + \frac{1}{2} - \text{round} (4i + 2)$$

$$= 4i + \frac{5}{2} - \text{round} (4i + 2)$$

WHERE FOR V

$$\begin{cases} V = q - (l + 0.5) & l = \text{round}(q) - 1 \\ q = \frac{j + 0.5}{0.25} \end{cases}$$

$$\begin{aligned} V &= (j + 0.5)4 - l - 0.5 = (j + 0.5)4 - \text{round}(q) + 1 - 0.5 \\ &= (j + 0.5)4 - \text{round}((j + 0.5)4) + \frac{1}{2} \\ &= 4j + 2 + \frac{1}{2} - \text{round}(4j + 2) \\ &= 4j + \frac{5}{2} - \text{round}(4j + 2) \end{aligned}$$

Let's test

$$4i + \frac{5}{2} - \text{round}(4i + 2)$$

$$i=0 \quad \frac{5}{2} - \text{round}(2)$$

$$i=1 \quad \frac{8+5}{2} - \frac{12}{2} = \frac{1}{2}$$

$$\frac{5}{2} - \frac{4}{2} = \frac{1}{2}$$

So both V & v are $\frac{1}{2}$.

$$\begin{aligned}
\tilde{\varphi}_{i,j}^2 &= (1-u) \left((1-v) \tilde{\varphi}_{k,l}^1 + v \tilde{\varphi}_{k,l+1}^1 \right) + \\
&\quad | \quad u \left((1-v) \tilde{\varphi}_{k+1,l}^1 + v \tilde{\varphi}_{k+1,l+1}^1 \right) \\
&= \left(\frac{1}{2}\right) \left(\left(\frac{1}{2}\right) \tilde{\varphi}_{k,l}^1 + \left(\frac{1}{2}\right) \tilde{\varphi}_{k,l+1}^1 \right) + \left(\frac{1}{2}\right) \left(\left(\frac{1}{2}\right) \tilde{\varphi}_{k+1,l}^1 + \left(\frac{1}{2}\right) \tilde{\varphi}_{k+1,l+1}^1 \right) \\
&= \left(\frac{1}{2}\right) \left(\left(\frac{1}{2}\right) \tilde{\varphi}_{k,l}^1 + \left(\frac{1}{2}\right) \tilde{\varphi}_{k,l+1}^1 + \left(\frac{1}{2}\right) \tilde{\varphi}_{k+1,l}^1 + \left(\frac{1}{2}\right) \tilde{\varphi}_{k+1,l+1}^1 \right) \\
&= \left(\frac{1}{2}\right) \left(\frac{1}{2} \left(\tilde{\varphi}_{k,l}^1 + \tilde{\varphi}_{k,l+1}^1 + \tilde{\varphi}_{k+1,l}^1 + \tilde{\varphi}_{k+1,l+1}^1 \right) \right)
\end{aligned}$$

② CONVERT MAX SPACE TO PPI

$$W = 6 \text{ cm}$$

$$h = 5 \text{ cm}$$

max space in inches

$$W = \frac{6}{2.54}$$

$$h = \frac{5}{2.54}$$

max space in pixels

$$= \text{PPI} \cdot W \text{ inches}$$

$$h = \text{PPI} \cdot h \text{ inches}$$

③ SAMPLING & QUANTIZATION

For pixel i, j in image A (sampled)

$$x = (i + 0.5) \Delta H$$

$$y = (j + 0.5) \Delta W$$

$$\Delta H = \frac{H}{M} - \text{unsampled height}$$

\sim sampled height

$$f_{ij} = \underbrace{f(x, y)}_{\text{SAMPLING}}$$

intensity in
original image

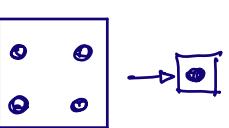
$$\Delta W = \frac{W}{N} - \text{unsampled width}$$

\sim sampled width

$$a_{ij} = \lfloor L \cdot f_{ij} \rfloor \quad \begin{matrix} L = \text{intensity levels} \\ f_{ij} = \text{sampled result} \end{matrix}$$

(QUANTIZATION)

④ SUPERSAMPLING

-  \rightarrow  $\frac{1}{4} \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}$ multiple points mean in one pixel

- Spatial Integration

$$f_{ij} = \frac{1}{\Delta W \cdot \Delta H} \int_{j \Delta W}^{(j+1) \Delta W} \int_{i \Delta H}^{(i+1) \Delta H} f(x, y) dx dy$$

⑤ QUANTIZATION ISSUES AND SOLUTIONS

Basic: $\hat{d}_{ij} = \lfloor L \cdot f_{ij} \rfloor$ with range $0, \dots, L-1$

\Rightarrow IF f_{ij} is at max value $\Rightarrow \hat{d}_{ij} = L$ OUT OF RANGE

Sol 1: $\hat{d}_{ij} = \text{round}\left(L f_{ij} - 0.5\right)$

\Rightarrow Works only if $\text{round}(0.5) = 0$

Sol 2: $\hat{d}_{ij} = \text{round}\left((L-1) f_{ij}\right)$

\Rightarrow Now in range, but max bucket $L-1$ is smaller!

Sol 3: $\hat{d}_{ij} = \left\lfloor L \cdot \frac{f_{ij} - f_*}{f^* - f_*} \right\rfloor$ where

$$f_* = \min_{x,y} f_{x,y}$$

$$f^* = \max_{x,y} f_{x,y}$$

⑥ BILINEAR INTERPOLATION

For ∂^2_{ij} intensity:

$$x = (i + 0.5) \cdot [\text{IMAGE MODIFICATION}]$$

$$y = (j + 0.5) \cdot [\text{IMAGE MODIFICATION}]$$

$$k = \text{round}(x) - 1$$

$$l = \text{round}(y) - 1$$

$$u = x - (k + 0.5)$$

$$v = y - (l + 0.5)$$

$$\begin{aligned} \partial^2_{ij} &= \text{round} \left((1-u)(1-v)\partial^1_{k,l} + (1-u)v\partial^1_{k+1,l} \right. \\ &\quad \left. + uv(1-v)\partial^1_{k,l+1} + uv(0)\partial^1_{k+1,l+1} \right) \end{aligned}$$

WITH INTEGRAL

$$\partial^2_{ij} = \int_{y-0.5}^{y+0.5} \int_{x-0.5}^{x+0.5} f_1(s,t) ds dt$$

⑦ GEOMETRIC TRANSFORMATIONS

- SCALING $\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} C_x & 0 \\ 0 & C_y \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = C \begin{pmatrix} x \\ y \end{pmatrix}$

- TRANSLATION $\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} x + t_x \\ y + t_y \end{pmatrix}$

- ROTATION $\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = R \begin{pmatrix} x \\ y \end{pmatrix}$

- HORIZONTAL SHEAR $\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} x \\ y + S_h x \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ S_h & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = S_h \begin{pmatrix} x \\ y \end{pmatrix}$

- VERTICAL SHEAR $\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} x + S_v y \\ y \end{pmatrix} = \begin{pmatrix} 1 & S_v \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = S_v \begin{pmatrix} x \\ y \end{pmatrix}$

⑧ HOMOGENEOUS COORDINATES TRANSFORMATIONS (AFFINE)

- INVARIANT TO AFFINE COMBINATIONS

- SCALING

$$C(c_x, c_y) = \begin{pmatrix} c_x & 0 & 0 \\ 0 & c_y & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

- INVERSE

$$C(c_x, c_y)^{-1} = C\left(\frac{1}{c_x}, \frac{1}{c_y}\right)$$

- TRANSLATION

$$T(t_x, t_y) = \begin{pmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \\ 0 & 0 & 1 \end{pmatrix}$$

- INVERSE

$$T(t_x, t_y)^{-1} = T(-t_x, -t_y)$$

- ROTATION

$$R(\theta) = \begin{pmatrix} \cos\theta & -\sin\theta & 0 \\ \sin\theta & \cos\theta & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

- INVERSE

$$R(\theta)^{-1} = R(-\theta)$$

- VERTICAL SHEAR

$$S_v(\lambda) = \begin{pmatrix} 1 & \lambda & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$S_v(\lambda)^{-1} = S_v(-\lambda)$$

- HORIZONTAL SHEAR

$$S_h(\lambda) = \begin{pmatrix} 1 & 0 & 0 \\ \lambda & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$S_h(\lambda)^{-1} = S_h(-\lambda)$$

⑨ FORWARD MAPPING

» For all pixels (x, y) of A_1 use TRANSFORM to compute (x', y') and transfer intensity.

ISSUES:

- Same pixel A_2 can have an A_1 mapping.
- Some pixel A_2 can lack a pre-image in A_1 .

⑩ INVERSE MAPPING

» For all pixels (x', y') in A_2 , use TRANSFORM^{-1} to compute (x, y) in A_1 and use any interpolation method to revert back to original.

⑪ 3-SHEARS ROTATION

$$R(\theta) = S_v(\lambda) S_h(\nu) S_v(\lambda)$$

$$\lambda = -\tan \frac{\theta}{2} \quad \nu = \sin \theta$$

(12)

CHANGE INTENSITY RESOLUTION

$$r \rightarrow \text{round} \left((r + 0.5) \frac{L_2}{L_1} - 0.5 \right)$$

(13)

TEMPORAL INTENSITY SUPER-SAMPLING

>> Sample source k times

$$\bar{A}_k = \frac{1}{k} \sum_{i=1}^k A_i; \quad k \rightarrow \infty \text{ No noise!}$$

>> On-line (Intensity formulation)

$$\bar{A}_{k+1} = \frac{k \cdot \bar{A}_k + A_{k+1}}{k+1}$$

>> IF UNCORRELATED NOISE

$$\sigma_{\bar{A}_k} = \frac{1}{\sqrt{k}} \sigma_n$$

(14) INTENSITY TRANSFORMATIONS

Discrete intensity transformation

$$\hat{T} : \{0, \dots, L_1 - 1\} : \{0, \dots, L_2 - 1\}$$

$$\hat{T}(r) = \text{round} \left(L \cdot \left(\frac{r + 0.5}{L} \right) - 0.5 \right)$$

- NEGATIVE

$$T(r) = 1 - r$$

- LOG TRANSFORM (enhance contrast in dark areas)

$$T(r) = \alpha \log \left(r / \beta + 1 \right)$$

- $\alpha = 1 / (\log \left(\frac{1}{\beta} + 1 \right))$
- $\beta = 1 / (\exp \left(\frac{1}{\alpha} \right) - 1)$

- EXP TRANSFORM (enhance contrast in bright areas)

$$T(r) = \beta \left(\exp \left(r / \alpha \right) - 1 \right)$$

- $\alpha = 1 / (\log \left(\frac{1}{\beta} + 1 \right))$
- $\beta = 1 / (\exp \left(\frac{1}{\alpha} \right) - 1)$

- GAMMA CORRECTION

$$T(r) = r^\gamma \quad \gamma > 0$$

if $\gamma < 1$ enhance in dark

if $\gamma > 1$ enhance in light

- CONTRAST STRETCHING

Medium intensity contrast enhance

$$T(r) = \frac{1}{2} \left(\frac{\arctan(\delta(2r-1)) + 1}{\arctan(\delta)} \right) \quad \delta > 0$$

- MORE -

$$T(r) = \frac{1}{2} \left(\lambda \frac{\arctan(\delta(2r-1)) + 1}{\arctan(\delta)} \right) \quad \delta > 0 \quad \lambda \geq 1$$

- THREE PIECES

$$S_1 = 0$$

$$S_S = 1$$

$$r_1 = \text{thresh1}$$

$$r_2 = \text{thresh2}$$

highlight one area

- INTENSITY LEVEL SLICING

$T(r)$ = piecewise
bijective
linear

(15) HISTOGRAM EQUALIZATION

$$\hat{T}(k) = \text{round}(\tilde{s}_k)$$

where

$$\tilde{s}_k = L \sum_{j=0}^{k-1} p_j + \frac{L}{2} p_k - \frac{1}{2} \quad \text{for } k=0, \dots, L-1$$

SANITY CHECK:

$$h_k = MN/L \text{ for } k=0, \dots, L-1$$

$$p_k = \frac{1}{L} \quad \tilde{s}_k = k \quad \hat{T}(k) = k \checkmark$$

(16) HISTOGRAM MATCHING

Find $\circ T$ transformation, such that the probability density function of g matches some target $q(s)$

$$T = T_q^{-1} \circ T_p \quad T_p(r) = \int_0^r p(z) dz \quad T_q(s) = \int_0^s q(z) dz$$

IN DISCRETE SETTING

$$T_p \text{ linear with } T_p(k/L) = \bar{p}_k \quad \bar{p}_k = \sum_{j=0}^{k-1} p_j$$

$$T_q \text{ linear with } T_q(k/L) = \bar{q}_k \quad \bar{q}_k = \sum_{j=0}^{k-1} q_j$$

$$T_q^{-1} \text{ linear with } T_q^{-1}(\bar{q}_k) = k/L$$

(17) LINEAR SPATIAL FILTERING

- box $\sigma = b$, $m = n = 2\sigma + 1$, $W_{S,f} = \frac{1}{m^2}$

- gaussian $\sigma = b$

$$W_{S,f} = \frac{\exp\left(-\frac{s^2+t^2}{2\sigma^2}\right)}{-2\left(\exp\left(-\frac{s^2+t^2}{2\sigma^2}\right)\right)}$$

Padding
 (zero
 mirror
 replicate)

PERFORMANCE : $O(MNmn)$

(18) MEDIAN FILTER

Get median value of kernel window

(19) LAPLACIAN FILTER (smoothing)

$$\text{OUT} = \underline{l} \star A$$

$\underline{l} = \begin{bmatrix} 0 & 1 & 0 \\ 1 & -4 & 1 \\ 0 & 1 & 0 \end{bmatrix}$ or $\underline{l}' = \begin{bmatrix} 0 & 1 & 0 \\ 1 & -8 & 1 \\ 0 & 1 & 0 \end{bmatrix}$
 passing filter

②₀ PREWITT FILTER (Box smoothing)

$$OUT = \sqrt{(g_x * A)^2 + (g_y * A)^2}$$

where:

$$g_x = \begin{bmatrix} -1 & -1 & -1 \\ 0 & 0 & 0 \\ 1 & 1 & 1 \end{bmatrix} \quad \text{and} \quad g_y = \begin{bmatrix} -1 & 0 & 1 \\ -1 & 0 & 1 \\ -1 & 0 & 1 \end{bmatrix}$$

②₁ SOBEL FILTER

$$OUT = \sqrt{(g_x * A)^2 + (g_y * A)^2}$$

where:

$$g_x = \begin{bmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ 1 & 2 & 1 \end{bmatrix} \quad \text{and} \quad g_y = \begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix}$$

②₂ BASIC BINARIZATION

$$\left\{ \begin{array}{ll} 0 & \text{if } A \leq t \\ 1 & \text{otherwise} \end{array} \right. \quad \parallel \quad \begin{array}{l} \text{threshold is selected by} \\ \text{fitting a poly OR a gauss.} \end{array}$$

(23) IMAGE SEGMENTATION BASICS

- conditions:
- $\bigcup_{i=1}^n R_i = R$
 - R_i connected set $\forall i \in 0, \dots, n$
 - $R_i \cap R_j = \emptyset$ for all i, j
 - IF SAME VALUES BUT NOT CONNECTED
 \hookrightarrow Different regions

(24) SIGNAL SMOOTHING $\stackrel{?}{\in}$ AVG. SMOOTHING

n = kernel size

$$\underline{\text{Signal Out}} \quad g_i = \frac{1}{n} \int_{i-\frac{n}{2}}^{i+\frac{n}{2}} f(x) dx$$

$n \times m$ = kernel size

$$B_{ij} = \frac{1}{n \cdot m} \int_{i-\frac{n}{2}}^{i+\frac{n}{2}} \int_{j-\frac{m}{2}}^{j+\frac{m}{2}} A(s, t) dt ds$$

$$B_{ij} = \sum_{s=-\frac{n}{2}}^{\frac{n}{2}} \sum_{t=-\frac{m}{2}}^{\frac{m}{2}} A_{ij} w(i+s, j+t)$$

$$W = \begin{bmatrix} \frac{1}{n+m} & \cdots & \frac{1}{n+m} \\ \vdots & \ddots & \vdots \\ \frac{1}{n+m} & \cdots & \frac{1}{n+m} \end{bmatrix}$$

(25) EDGE DETECTION (BASIC CONCEPT)

$$g(x) = \begin{cases} 1 & \text{if } |f'(x)| > t \\ 0 & \text{otherwise} \end{cases}$$

edge detected/not

(26) EDGE DETECTION WITH FIRST ORDER DER.

$$W_x = \frac{\partial f(x, y)}{\partial x} = f'_x \approx \frac{f(x+1, y) - f(x-1, y)}{2}$$

$$W_y = \frac{\partial f(x, y)}{\partial y} = f'_y \approx \frac{f(x, y+1) - f(x, y-1)}{2}$$

$$W_x = \begin{bmatrix} -\frac{1}{2} & 0 & \frac{1}{2} \end{bmatrix} \quad W_y = \begin{bmatrix} 0 & \frac{1}{2} \\ 0 & 0 \\ \frac{1}{2} & 0 \end{bmatrix}$$

(27) EDGE PROPERTIES

- Position

$$(x, y)$$

$$W_x = \begin{bmatrix} -\frac{1}{2} & 0 & \frac{1}{2} \end{bmatrix} \quad W_y = \begin{bmatrix} -\frac{1}{2} \\ 0 \\ \frac{1}{2} \end{bmatrix}$$

- Magnitude

$$\left. \begin{array}{l} G_x = W_x * A \\ G_y = W_y * A \end{array} \right\} \Rightarrow \begin{array}{l} \text{Magnitude}_G = \sqrt{G_x^2 + G_y^2} \\ \cong |G_x| + |G_y| \end{array}$$

- Orientation

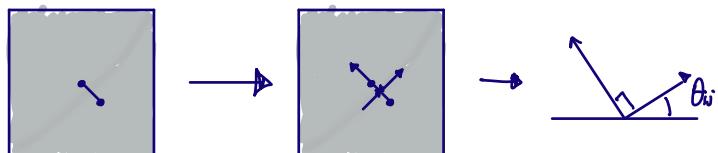
$$\Theta = \arctan \frac{G_x}{G_y}$$

(28) CANNY EDGE DETECTION

- INTEGRATION TO SMOOTH IMAGE
- DIFFERENTIATION (27)
- NON-MAXIMA SUPPRESSION: (29)
- HYSTERESIS (DUAL THRESHOLDING) (30)
- CONNECTIVITY ANALYSIS (31)

(29) NON-MAXIMA SUPPRESSION

Transform edge into a single line of pixels



A) Group angles: $\theta_i^j \begin{cases} 0 & \text{if } -22.5 \leq \theta_{ij} < 22.5 \\ 45 & \text{if } 22.5 \leq \theta_{ij} < 67.5 \\ 90 & \text{if } 67.5 \leq \theta_{ij} < 112.5 \\ 135 & \text{if } 112.5 \leq \theta_{ij} < 157.5 \end{cases}$

B) Compare magnitude + neighbours

$$\left| \begin{array}{l} \text{if } \theta_{ij}^j = 0: \\ \quad \text{if } (G_{ij} < G_{(i-1),j} \parallel G_{ij} < G_{(i+1),j}): \\ \quad \quad G_{ij}^1 = 0 \\ \text{if } \theta_{ij}^j = 45: \\ \quad \text{if } (G_{ij} < G_{(i-1),(j+1)} \parallel G_{ij} < G_{(i+1),(j-1)}): \\ \quad \quad G_{ij}^1 = 0 \end{array} \right| \quad \left| \begin{array}{l} \text{if } \theta_{ij}^j = 90: \\ \quad \text{if } (G_{ij} < G_{(i-1),j} \parallel G_{ij} < G_{(i+1),j}): \\ \quad \quad G_{ij}^1 = 0 \\ \text{if } \theta_{ij}^j = 135: \\ \quad \text{if } (G_{ij} < G_{(i-1),j} \parallel G_{ij} < G_{(i+1),j}): \\ \quad \quad G_{ij}^1 = 0 \\ \text{else:} \\ \quad \quad G_{ij}^1 = G_{ij} \end{array} \right.$$

(30)

HYSTERESIS THRESHOLDING

Set low & high threshold

pixels above t_{high} form the image G'_{high}

pixels above t_{low} form the image G'_{low}

G'_{high} will contain pixels of large magnitude edges

G'_{low} will contain pixels of low magnitude edges

$$G''_{i,j} = \begin{cases} 2 & \text{if } G'_{i,j} \geq t_{high} \\ 1 & \text{if } t_{high} > G'_{i,j} \geq t_{low} \\ 0 & \text{otherwise} \end{cases}$$

(31)

CONNECTIVITY ANALYSIS

Use result of (30) $\rightarrow G''_{i,j}$ to compute an edge image.

- Region growing :
 - Get a "seed" pixel with $G''_{i,j} = \underline{\text{class 2}}$
 - Grow seed including neighbours with magnitude ≥ 1
 - Get new seed with $G''_{i,j} = \underline{\text{class 2}}$
 - repeat until all $G''_{i,j} = 2$ are included

(32)

HOUGH TRANSFORM HORIZONTAL LINES

- Slope-intercept equation: $y_i = \alpha x_i + b$

with $\alpha = \underline{0}$

Horizontal!

$$\boxed{y_i = b}$$

- Create accumulator of discrete voting bins where each bin is a row in H

$$A \text{ 120 rows} \Rightarrow |H| = 120$$

For each (x_i, y_i) in A , we take the position of the x axis and add to $H[x_i]$ the intensity of δ_{x_i, y_i} .

We then apply non-maxima suppression (29)

(33)

HOUGH TRANSFORM ON LINES ON ANY CHOSEN ANGLE α

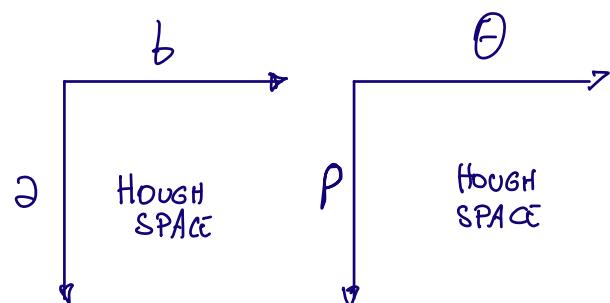
- compute orientation of gradients (27) θ and keep only points with θ close to the angle α to use

- Then put intensities in Hough space following α and b of line equation:

$$y_i = \alpha x_i + b$$

or

$$x_i \cos \theta + y_i \sin \theta = P$$

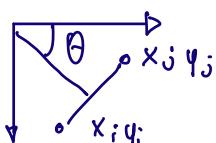


(34)

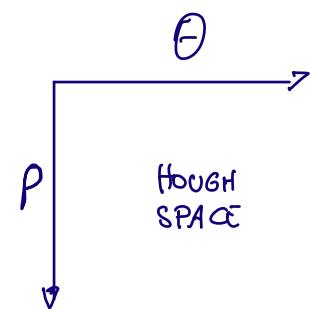
LINE HOUGH TRANSFORM (COMPLETE)

Use normal representation of line using:

$$(x_i \cos \theta + y_i \sin \theta = P)$$



and use accumulator:



(35) HOUGH CIRCLES

use same procedure but with a, b
and a chosen r

$$(x_i - a)^2 + (y_i - b)^2 = r^2$$

(36) HARRIS - STEPHENS CORNER DETECTOR

Using
Kernel = $m \times n$

- SUM OF SQUARE DIFFERENCES AT POSITION (s, t)

$$C(x, y) = \sum_{S=-\frac{n}{2}}^{\frac{n}{2}} \sum_{T=-\frac{m}{2}}^{\frac{m}{2}} [A(s+x, t+y) - A(s, t)]^2$$

- WEIGHTED SUM OF SQUARE DIFFERENCES AT POSITION

$$C(x, y) = \sum_{S=-\frac{n}{2}}^{\frac{n}{2}} \sum_{T=-\frac{m}{2}}^{\frac{m}{2}} W(st) \underbrace{[A(s+x, t+y) - A(s, t)]^2}_{\text{SMOOTHING KERNEL } (s, t)}$$

- BILINEAR APPROXIMATION

$$C(x, y) = \sum_{s=-\frac{n}{2}}^{\frac{n}{2}} \sum_{t=-\frac{m}{2}}^{\frac{m}{2}} W(st) \left[\left(A(s, t) + x G_x(s, t) + y G_y(s, t) \right) - A(s, t) \right]^2$$

$$C(x, y) = [x \ y] M \begin{bmatrix} x \\ y \end{bmatrix}$$

with $M = \sum_s \sum_t W(s, t) \begin{bmatrix} G_x(s, t)^2 & G_x(s, t) G_y(s, t) \\ G_x(s, t) G_y(s, t) & G_y(s, t)^2 \end{bmatrix}$

Harris matrix

where $G_x = A \star \begin{bmatrix} -0.5 \\ 0 \\ 0.5 \end{bmatrix}$ $G_y = A \star \begin{bmatrix} -0.5 & 0 & 0.5 \end{bmatrix}$

(37) HARRIS-STEVENS WITH EIGENVAL

CALCULATE $\lambda_x \notin \lambda_y$ of M (36) and

use that to find $\det(M) = \lambda_x + \lambda_y$

$$\text{trace}(M) = \lambda_x + \lambda_y$$

$$S: R = \det(M) - K \cdot (\text{trace}(M))^2$$

$$\det(M) = G_x(s,t)^2 G_y(s,t)^2$$

$$\text{trace}(M) = G_x(s,t)^2 + G_y(s,t)^2$$

$$K = [0.04, 0.06]$$

(38) COLOR XYZ

$$X = \int \bar{x}(\lambda) I(\lambda) d\lambda \quad \leftarrow L \text{ cone}$$

$$Y = \int \bar{y}(\lambda) I(\lambda) d\lambda \quad \leftarrow M \text{ cone}$$

$$Z = \int \bar{z}(\lambda) I(\lambda) d\lambda \quad \leftarrow S \text{ cone}$$

λ : SPD Light

(39) COLOR $x y Y$

$$X = \frac{X}{X+Y+Z} \quad Y = \frac{Y}{X+Y+Z} \quad Z = \frac{Z}{X+Y+Z} = 1-X-Y$$

We Just use

x, y and $\begin{matrix} Y \\ | \\ \text{Luminance} \end{matrix}$

(40) RGB \leftrightarrow XYZ

$$\begin{pmatrix} R \\ G \\ B \end{pmatrix} = \begin{pmatrix} 3.06 & -1.39 & -0.48 \\ -0.97 & 1.88 & 0.04 \\ 0.07 & -0.23 & 1.07 \end{pmatrix} \begin{pmatrix} X \\ Y \\ Z \end{pmatrix}$$

(41) RGB \leftrightarrow CMY

$$\begin{pmatrix} C \\ M \\ Y \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} - \begin{pmatrix} R \\ G \\ B \end{pmatrix} \quad \begin{pmatrix} R \\ G \\ B \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} - \begin{pmatrix} C \\ M \\ Y \end{pmatrix}$$

(42) CMY \leftrightarrow CMYK

$$(CMY) = \frac{(C-K, M-K, Y-K)}{(1-K)}$$

(43) RGB \leftrightarrow HSI

$$H = \begin{cases} \theta & \text{if } B \leq G \\ 360 - \theta & \text{if } B > G \end{cases} \quad \theta = \arcsin \left(\frac{\frac{1}{2} [(R-G) + (R-B)]}{\sqrt{(R-G)^2 + (R-G)(G-B)}} \right)$$

$$S = 1 - \frac{3 \min(R, G, B)}{R + G + B} = 1 - \frac{\min(R, G, B)}{I}$$

$$I = \frac{R + G + B}{3}$$

(44) HSI \leftrightarrow RGB

$$\text{if } H \in [0, 120)$$

$$B = I(1-s)$$

$$R = I \left(1 - \frac{S \cos H}{\cos(60-H)} \right)$$

$$G = 3I - (R+B)$$

$$\text{elif } H \in [120, 240)$$

$$H = H - 120$$

$$R = I(1-s)$$

$$G = I \left(1 - \frac{S \cos H}{\cos(60-H)} \right)$$

$$B = 3I - (R+G)$$

$$\text{elif } H \in [240, 360)$$

$$H = H - 240$$

$$G = I(1-s)$$

$$B = I \left(1 - \frac{S \cos 60}{\cos(60-H)} \right)$$

$$R = 3I - (G+B)$$

(45) PSEUDO - COLORS

in $f: X \times Y \rightarrow I$ $T_R, T_G, T_B \rightarrow I$

out: $g_R, g_G, g_B: X \times Y \rightarrow I$

where $g \star (x, y) = T \star (f(x, y))$

(46) CROSS-CORRELATION vs CONVOLUTION

CROSS-CORR

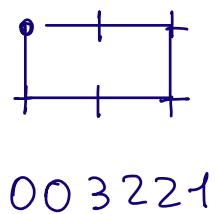
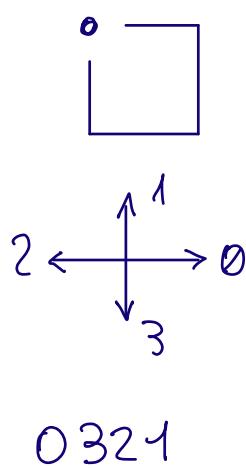
$$\begin{matrix} & \star \\ \begin{matrix} x \\ y \\ z \end{matrix} & \begin{bmatrix} A & B & C \\ D & E & F \\ G & H & I \end{bmatrix} \end{matrix} = \boxed{\begin{matrix} \hat{x} & \hat{y} & \hat{z} \\ i & j & n \\ w & l & m \end{matrix}} \quad Ax + By + Cz + jD + \dots$$

CONVOLUTION

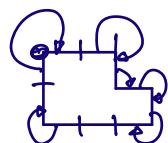
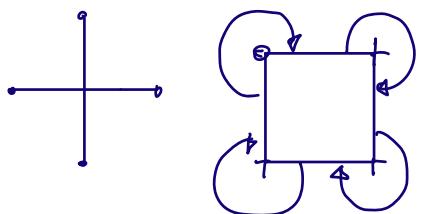
$$\begin{matrix} & \star \\ \begin{matrix} x \\ y \\ z \end{matrix} & \begin{bmatrix} A & B & C \\ D & E & F \\ G & H & I \end{bmatrix} \end{matrix} = \boxed{\begin{matrix} \hat{x} & \hat{y} & \hat{z} \\ i & j & n \\ w & l & m \end{matrix}} \quad I*x + H*y + G*z + \dots$$

(47) STRUCTURAL PATTERN MATCHING

- SHAPE NUMBERS



* - SHAPE NUMBERS (Rotation Inv.)

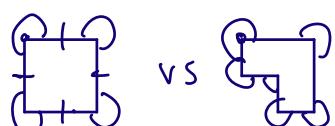


- STRING-BASED PATTERN MATCHING

>> SHIFT THE RESULT OF * FOR BOTH IMAGES

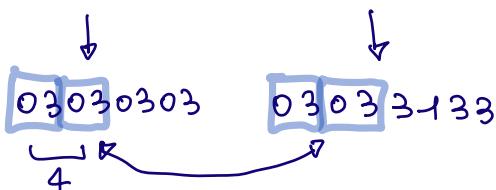
TO GET THE MOST 0s on the left

And check result 2 at a time



$$\text{So } \frac{4}{8} = \frac{1}{2} \quad \text{DISTANCE} = 1 - \frac{1}{2}$$

30303030 30303313



(48) CONIC SECTION :

$$\{(x,y)^T \in \mathbb{R}^2 : ax^2 + bxy + cy^2 + dx + ey + f = 0\}$$

$$\text{DISCRIMINANT: } \Delta = b^2 - 4ac$$

$$\Delta = -4\Delta$$

$$\text{where } \Delta = \det M \quad \text{where } M = \begin{pmatrix} a & b/2 \\ b/2 & c \end{pmatrix}$$

$$\text{so } \frac{\Delta}{\text{DISCRIMINANT}} = -4 \det \begin{pmatrix} a & b/2 \\ b/2 & c \end{pmatrix}$$

(49) CONIC TYPES BASED ON DETERMINANT

$$\left. \begin{array}{l} \Delta > 0 : \text{hyperbola} \\ \Delta < 0 : \text{ellipse} \\ \Delta = 0 : \text{parabola} \end{array} \right\} \begin{array}{l} \text{if } a, b, c \text{ are} \\ \text{not all equal to 0} \end{array}$$



(50) CONIC IN \mathbb{P}^2 HOMOGENEOUS

$$ax^2 + bxy + cy^2 + dxz + eyz + fz^2 = 0$$

if z is not equal to 1:

To prove this replace $x \rightarrow zx$ and $y \rightarrow zy$

then

$$az^2 + bxyz^2 + cy^2z^2 + dxz^2 + eyz^2 + fz^2 = 0$$

and if we devide all by z^2 to get to original.

- AS A QUADRATIC FORM: $M \Rightarrow$ it's determinant will be used for δ !

$$\boxed{x^T C x} = 0, \quad C = \begin{pmatrix} a & b/2 & d/c \\ b/2 & c & e/2 \\ d/c & e/2 & f \end{pmatrix}$$

point in homogeneous

$$x = (x, y, z)^T$$

$$\boxed{x^T C x} = 0$$

C is SYMMETRIC AND QUADRATIC FORM IF set to 0 we get all the points in HOMO COORDS

$[C] = \{\lambda C\}$ is still the same conic!

(S) CONICS & DEGREES OF FREEDOM

- Only 5 degrees of Freedom

→ so -

- CONIC CAN BE DEFINED UNIQUELY by
5 points $x_i = (x_i, y_i, z_i)^T$
 $i=1, \dots, 5$ in general position
not 3 of them collinear!

Why? each point x_i
gives one constraint
on the coefficients

$$\underbrace{(x_i^2, x_i y_i, y_i^2, x_i z_i, y_i z_i, z_i^2)}_{*} (a, b, c, d, e, f)^T = 0$$

* For each of the 5 defining points

>> IF WE ARRANGE THE 5 * AS A MATRIX,
WE GET X

$$\text{so: } Xc = 0$$

$$\begin{matrix} \text{Point 1:} \\ \text{Point 2:} \\ \vdots \\ \text{Point 5:} \end{matrix} \left[\begin{matrix} x_1^2, x_1 y_1, y_1^2, x_1 z_1, y_1 z_1, z_1^2 \\ x_2^2, x_2 y_2, y_2^2, x_2 z_2, y_2 z_2, z_2^2 \\ \vdots \\ x_5^2, x_5 y_5, y_5^2, x_5 z_5, y_5 z_5, z_5^2 \end{matrix} \right] \left[\begin{matrix} a \\ b \\ c \\ d \\ e \\ f \end{matrix} \right] = 0$$

"c must be in the null space of X "

(52) DEGENERATE CONIC

If 3 of the 5 DEFINING POINTS ARE
COLLINEAR \Rightarrow Degenerate

$\det(C) = 0$ will mean a rank-deficient
matrix
 \Downarrow
DEGENERATE CONIC

- If a conic is degenerate: there can be
more than one solution! more than 1
conic.

(53) DEGENERATE CONICS IDENTIFICATION

$S_{\text{DISCRIM.}} = \emptyset$: if $\text{rank}(C) = 1$ or $(d^2 + e^2 = 4(a+c)f)$:
one line
if $(d^2 + e^2 < 4(a+c)f)$:
complex conjugate lines [you don't see them]
Points with imaginary
else: 2 lines

$S_{\text{DISCRIM.}} < 0$: Single point (degenerate ellipse)
so an intersection of 2 non-parallel
complex conjugate lines

$S_{\text{DISCRIM.}} > 0$: Two intersecting lines (degenerate hyperbola)

(54)

PROOF OF $C = l m^T + m l^T$ IS THE
DEGENERATE CONIC COMPOSED OF
LINES l and m

• PROOVE ALL POINTS ON CONIC

A • IF:

$$\begin{array}{c} X \text{ is on } l \\ | \\ \text{column vect} \end{array} \Rightarrow x^T l = l^T x = 0$$

B • IF:

$$\begin{array}{c} X \text{ is on } m \\ | \\ \text{column vect} \end{array} \Rightarrow x^T m = m^T x = 0$$

IN BOTH A or B WE CAN STATE:

$$\begin{aligned} x^T C x &= x^T (l m^T + m l^T) x = (x^T l)(m^T x) + (x^T m)(l^T x) \\ l m^T + m l^T &= (x^T l)(m^T x) + (x^T m)(l^T x) \\ &\quad \swarrow \quad \searrow \\ &\quad \text{IF } x \text{ is on } m \quad \text{IF } x \text{ is on } l \\ &= 0 \end{aligned}$$

SO $x^T C x = 0$! SO x is on conic C So all points on l and all points on m are
on the conic C .

• PROOVE DEGENERATE

Now BECAUSE C is symmetric \Rightarrow As all ConicsTHEN: $C^T = C$

$$\downarrow \\ (l m^T + m l^T)^T = l m^T + m l^T \quad \left. \right\}$$

AND

Rank(C) = 2 WHY? BECAUSE Null space of
 C are all the multiples of x

$$\text{null}(C) = \left\{ \lambda x : \lambda \in \mathbb{R} \right\} \quad \begin{array}{c} x = l \times m \\ \text{intersection of} \\ \text{the 2 lines} \end{array}$$

Proof:

$$C(\lambda x) = (l m^T + m l^T)(\lambda x) = \lambda l m^T x + \lambda m l^T x = 0$$

x is on
line l and also
on line m as
before!

SO AS Rank-Deficient
Rank(C) = 2 \Rightarrow Degenerate
Conic

(55) TANGENT LINES TO CONIC (PROOF)

$\ell = Cx$
 |
 l point on the conic
Prove line tangent to conic C at x

Proof:

if ℓ contains x , then $x^T \ell = 0$

and then as $\ell = Cx$:

$x^T \underbrace{Cx}_{\ell} = 0 \Rightarrow \ell = Cx$ is \Rightarrow line through x

THEN:

- IF ℓ intersects C in 1 point
 then we have tangent! ✓

- OTHERWISE: we have point $y \neq x$ on ℓ
 and on C , then also

We check any point z on line ℓ $y^T C y = 0$ and $\ell^T y = 0$

$$(2x + \beta y)^T C (2x + \beta y) = \underbrace{2^T C x}_{\in \ell} + 2 \underbrace{\beta x^T C y}_{\in \ell} + \underbrace{\beta^2 y^T C y}_{\in \ell} = 0$$

so then degenerate! WE DON'T CONSIDER IT! ✓

(56) ADJOINT OF A CONIC

$$\text{adj}(C) = \begin{cases} \text{if } C \text{ is full rank then} \\ \text{adj}(C) = C^{-1} = C^* \end{cases}$$

- ℓ is tangent to C and an element
 of dual conic $C^* \Leftrightarrow \ell^T C^* \ell = 0$

(57) Proof that $\ell^T C^* \ell = 0$ CAPTURES ALL LINES IN C^*

- if x is on C , ex: $x^T C x = 0$ then $\ell = Cx$
is \Rightarrow line in C^*

- if C is full rank $\Rightarrow X = C^{-1} \ell$

$$\hookrightarrow \ell^T C^* \ell = \ell^T C^{-1} \ell = \ell^T C^{-T} C C^{-1} \ell$$

↓ Symmetry . I

$$= (C^{-1} \ell)^T C (C^{-1} \ell) = X^T C X = 0$$

(58) DESCRIBE P^n

- Points and lines in \mathbb{P}^2 are represented as vectors in $\mathbb{R}^3 \setminus \{0\}$

- Conics in \mathbb{P}^2 are represented as symmetric matrices from $\mathbb{R}^{3 \times 3} \setminus \{0\}$

(59) PROJECTIVE TRANSFORMATIONS IN 2D

Projective trans: mapping $h: \mathbb{P}^2 \rightarrow \mathbb{P}$

- h is linear in homo coords

- h is \Rightarrow non-singular matrix $H \in \mathbb{R}^{3 \times 3}$
so $h(x) = Hx$

- H is homogenous \Rightarrow can be scaled by any λ without changing

- 8 degrees of freedom: 4 original \Rightarrow 4 new points correspondences

⑥ PROVE 3 COLLINEAR POINTS STAY COLLINEAR
UNDER A PROJECTIVE TRANSFORMATION

» if x_i points are collinear $\Rightarrow \exists l$ line such that $x_i^T l = \emptyset$ for all points i

» if we take the transformed points

$$Hx_i \quad \begin{matrix} \text{we claim} \\ \text{they lie on the} \\ \text{transformed} \\ \text{line} \end{matrix} : \underbrace{(Hx_i)^T}_{x'_i} \underbrace{(H^{-T}l)}_{l'} = \emptyset \quad \begin{matrix} \text{we claim} \\ \text{transformed} \\ \text{points} \\ \text{are on} \\ \text{the line} \end{matrix}$$

Live transformed
with H^{-T}

$$\underbrace{(Hx_i)^T}_{x'_i} \underbrace{(H^{-T}l)}_{l'} = x_i^T \underbrace{H^T H^{-T} l}_{(H^{-1}H)^T} = x_i^T l = \emptyset$$

$$x'_i = Hx_i \quad \text{lie on } l' \stackrel{I}{=} \quad \checkmark$$

61 FIND MAPPING WITH H 4 points \Rightarrow 4 points

STARTING WITH 4 points: $\begin{matrix} x_1 u_1 & x_3 u_3 \\ x_2 u_2 & x_4 u_4 \end{matrix}$

CONVERT TO HOMOGENIUS $\begin{matrix} \text{Original}_1 = (x_1, u_1, 1) \\ \text{Original}_2 = (x_2, u_2, 1) \\ \text{Original}_3 = (x_3, u_3, 1) \\ \text{Original}_4 = (x_4, u_4, 1) \end{matrix}$

$$\frac{\text{new } X_1}{\text{new } Z_1} = \frac{H \cdot \text{Original}_1 X}{H \cdot \text{Original}_1 Z} \quad \frac{\text{new } Y_1}{\text{new } Z_1} = \frac{H \cdot \text{Original}_1 Y}{H \cdot \text{Original}_1 Z}$$

$$\frac{\text{new } X_2}{\text{new } Z_2} = \frac{H \cdot \text{Original}_2 X}{H \cdot \text{Original}_2 Z} \quad \frac{\text{new } Y_2}{\text{new } Z_2} = \frac{H \cdot \text{Original}_2 Y}{H \cdot \text{Original}_2 Z}$$

$$\frac{\text{new } X_3}{\text{new } Z_3} = \frac{H \cdot \text{Original}_3 X}{H \cdot \text{Original}_3 Z} \quad \frac{\text{new } Y_3}{\text{new } Z_3} = \frac{H \cdot \text{Original}_3 Y}{H \cdot \text{Original}_3 Z}$$

$$\frac{\text{new } X_4}{\text{new } Z_4} = \frac{H \cdot \text{Original}_4 X}{H \cdot \text{Original}_4 Z} \quad \frac{\text{new } Y_4}{\text{new } Z_4} = \frac{H \cdot \text{Original}_4 Y}{H \cdot \text{Original}_4 Z}$$

SOLVE H: and get matrix coefficients

Hg is 1! Homo

(62) TRANSFORM A POINT x TO POINT x'

$$x \mapsto x' = Hx$$

(63) TRANSFORM A LINE $|$ TO $|'$

$$| \mapsto |' = H^{-T} | \quad \text{PROOF } \textcircled{ss}$$

(64) CONIC C TO CONIC C'

$$C \mapsto C' = H^{-T} CH^{-1}$$

$\underbrace{\quad}_{\text{matrix of conic}}$

x lies on C if only if $x^T C x = 0$

$$Hx \text{ lie on } C' = H^{-T} CH^{-1}$$

let's check:

$$(Hx)^T C' (Hx) \stackrel{?}{=} 0 \Rightarrow \underbrace{(x^T H^T)}_{\Rightarrow x^T C x} \underbrace{(H^{-T} CH^{-1})}_{I} \underbrace{(Hx)}_{I} \checkmark$$

(65)

DUAL CONIC TRANSFORM PROOF

$$C^* \mapsto C^{*\prime} = HC^* H^T$$

$$l^T C^* l = 0$$

$$(H^{-T} l)^T (H C^* H^T) (H^{-T} l)$$

$$(l^T H^{-1}) \underbrace{(H C^* H^T)}_{\stackrel{I}{=}} (H^{-T} l)$$

$$l^T C^* l = 0 \quad \checkmark \quad \stackrel{I}{=}$$

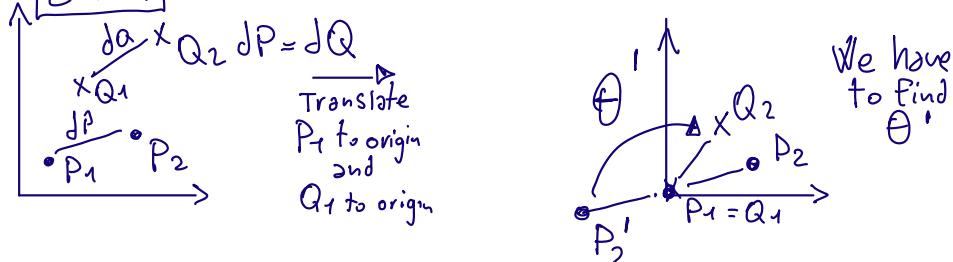
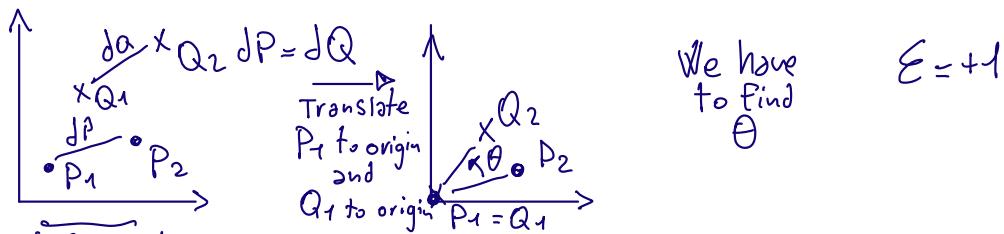
(66) SPECIAL CASE: ISOMETRIES

$$H_I = \begin{pmatrix} E \cos \theta & -\sin \theta & t_x \\ E \sin \theta & \cos \theta & t_y \\ 0 & 0 & 1 \end{pmatrix} \quad \text{IF } E = 1, -1$$

THEN WE GET
ROTATION + TRANSLATION
OR REFLECTION ($E = -1$)

3 DOF pws ϵ

- Determined by 2 points (equidistant pairs)



(67) EUCLIDEAN TRANSFORMATIONS (DISPLACEMENT)

Rotation + Translation

$$H_E = \begin{pmatrix} R_\theta & t_x \\ \begin{matrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{matrix} & t_y \\ 0 & 0 & 1 \end{pmatrix}$$

R_θ is orthogonal: $R_\theta^T R_\theta = R_\theta R_\theta^T = I$

IF $t=0$ pure rotation

IF $R_\theta = I$ pure translation

(68) SIMILARITIES (SHAPE-PRESERVING)

$$H_S = \begin{pmatrix} E\cos\theta & -E\sin\theta & t_x \\ E\sin\theta & E\cos\theta & t_y \\ 0 & 0 & 1 \end{pmatrix} \quad E = \pm 1$$

S = SCALING

4 DOF PLUS E

with 2 points correspondence

INVARIANT: angles, ratios of length

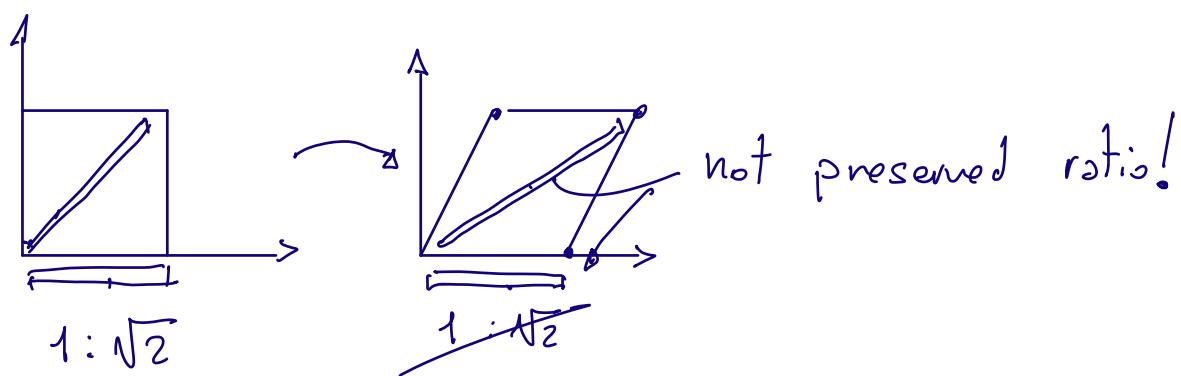
(69) AFFINE TRANSFORMATIONS

$$H_A = \begin{pmatrix} \alpha_{11} & \alpha_{12} & t_x \\ \alpha_{21} & \alpha_{22} & t_y \\ 0 & 0 & 1 \end{pmatrix} \quad A = \begin{pmatrix} \alpha_{11} & \alpha_{12} \\ \alpha_{21} & \alpha_{22} \end{pmatrix} \text{ non singular}$$

6 degs freedom 3-point correspondence

INVARIANT:

- Parallel lines and ideal points as $\ell_\infty = H_A^{-T} \ell_\infty$
- Ratios of lengths of parallel line segments
- Ratios of areas
- example of shear:



70) COMPACT REPRESENTATION OF AFFINE TRANS

$$H_A = \begin{pmatrix} A & + \\ 0^T & 1 \end{pmatrix}$$

IN CARTESIAN

$$\mathbf{x} = \begin{pmatrix} x \\ 1 \end{pmatrix} \mapsto Ax + t$$

How?

$$\mathbf{x} = \begin{pmatrix} x \\ 1 \end{pmatrix}$$

$$\left(\begin{pmatrix} A & + \\ 0^T & 1 \end{pmatrix} \right) \left(\begin{pmatrix} x \\ 1 \end{pmatrix} \right) = \left(\begin{pmatrix} Ax \\ 1 \end{pmatrix} + t \right) \rightsquigarrow Ax + t$$

71) DECOMPOSE LINEAR PART OF AFFINE TRANS
IN ROTATION AND SCALINGS

$$A = R\psi \underbrace{\begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}}_{\text{Scaling}} R\phi = R\theta R_{-\phi} \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} R\phi$$

SING. VALUE DECOMP: $A = UDV^T$ with $D = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$

$\downarrow \quad \downarrow$
 $R\psi \quad R\phi$

U, V orthogonal

$$A = \underbrace{UV^T}_{R\theta} \underbrace{VDV^T}_{R_{-\phi}} \underbrace{V^T}_{R\phi}$$

(72)

AFFINE MAPPING, SO HOW THAT INFINITY LINE MAPPED TO INFINITY LINE

$$H_A^{-T} \cdot l_\infty \stackrel{?}{=} l_\infty \quad \text{with } l_\infty = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$H_A = \begin{pmatrix} A & + \\ 0 & 1 \end{pmatrix}, \quad H_A^{-1} = \begin{pmatrix} A^{-1} & -A^{-1}+ \\ 0^T & 1 \end{pmatrix}$$

↓ Proof

$$H_A \cdot H_A^{-1} = \begin{pmatrix} I & -AA^{-1}+ + + = 0 \\ 0^T & 1 \end{pmatrix} = I_{3 \times 3}$$

Line at ∞

$$H_A^{-T} = \begin{pmatrix} A^{-T} & 0 \\ -f^T A^{-T} & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

Line at ∞ ✓

⑦3 PROJECTIVE TRANSFORMATION

(like affine but with anything in last row)

$$H_P = \begin{pmatrix} A & f \\ v^T & w \end{pmatrix} \quad \text{where } v^T = (v_1, v_2) \text{ } \left(\begin{matrix} v_1 \\ v_2 \end{matrix} \right) \text{ any row vector} \\ \text{with still } \det(H_P) \neq 0$$

IN CARTESIAN

$$x = \begin{pmatrix} x \\ y \\ 1 \end{pmatrix} \mapsto \frac{Ax + f}{v^T x + w}$$

Proof

$$\begin{pmatrix} A & f \\ v^T & w \end{pmatrix} \begin{pmatrix} x \\ y \\ 1 \end{pmatrix} = \begin{pmatrix} Ax + f \\ v^T x + w \\ 1 \end{pmatrix} \rightsquigarrow \underbrace{\frac{Ax + f}{v^T x + w}}_{\text{map at } \infty} \quad \text{if } v^T x + w = 0$$

ALSO TRANSFORMS COORDINATES AT ∞ !

$$\begin{pmatrix} A & f \\ v^T & w \end{pmatrix} \begin{pmatrix} x \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} Ax \\ v^T x \\ 0 \end{pmatrix} \rightsquigarrow \underbrace{\frac{Ax}{v^T x}}_{\text{if } v^T x = 0 \text{ THEY STAY AT } \infty} \quad \text{point at } \infty$$

IF $v=0$ and $w \neq 0 \rightarrow$ affine trans

• 8 DOF Determined by 4 points = No 3 collinear

INVARIANTS

- lives
- cross ratios $\frac{\|P_3 - P_1\| \cdot \|P_4 - P_2\|}{\|P_3 - P_2\| \cdot \|P_4 - P_1\|}$

⑦4 PROJECTIVE TRANSFORMATION LINE AT ∞

$$l_\infty = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \text{ maps to } l'_\infty = \begin{pmatrix} -A^{-T}v \\ 1 \end{pmatrix}$$

Proof:

$$\begin{pmatrix} A & -T \\ v^T w \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} -A^{-T}v \\ 1 \end{pmatrix}$$

So line at ∞
not necessarily
maps to a line
at ∞

except if $v = \emptyset$

⑦5 TRANS PROPERTIES (WHAT TO CHOOSE?)

- PROJECTIVE \rightarrow IN VARIANT:
 - Concurrency
 - Collinearity
 - Order of contact
(intersect, tangency,)
 - inflection
 - cross ratio
- AFFINE \rightarrow IN VARIANT:
 - parallelism, ratio of area, ratio of length on collinear, line at infinity
- SIMILARITY \rightarrow IN VARIANT:
 - ratio of lengths, angle, circular point
- EUCLIDEAN \rightarrow IN VARIANT:
 - length, area

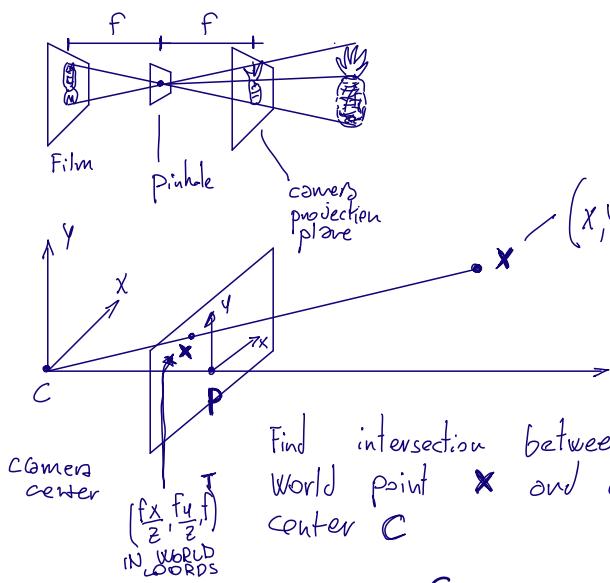
76) CAMERA BASICS

It's a projection from world point \mathbb{R}^3 to an image point \mathbb{R}^2 .

$$\cdot P = \mathbb{R}^{3 \times 4}$$

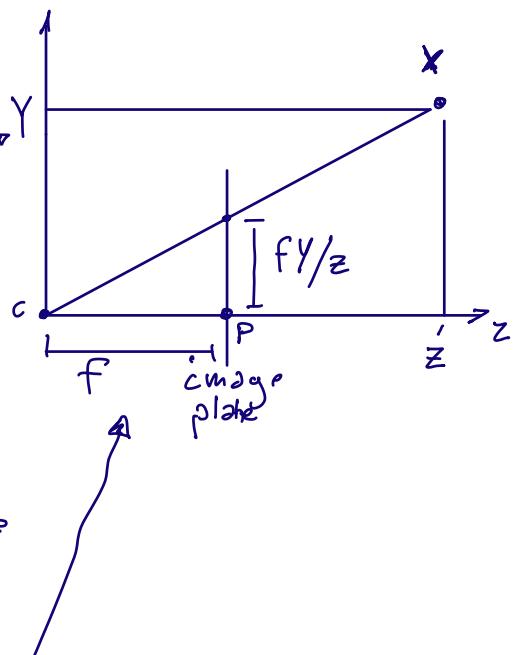
WHERE P has 11 DOF

77) PINHOLE CAMERA BASIC



Find intersection between world point X and camera center C

$$\frac{y}{y} = \frac{f}{z} \left| \begin{array}{l} \text{ratio of } z \text{ coordinate} \\ = \frac{f \cdot y}{z} \end{array} \right.$$



x coordinate is just f !

f is the same on all points, so we can ignore it

78 PINHOLE CAMERA MAPPING

$$\mathbf{X} \rightarrow \mathbf{x} = \mathbf{P} \mathbf{X}^{\text{real}}$$

| camera
mapping

FROM A WORLD POINT $\mathbf{X} = (x, y, z, 1)^T$ in \mathbb{R}^3

TO AN IMAGE POINT $\mathbf{x} = (f_x, f_y, f)^T$ in \mathbb{R}^2

USING \mathbf{P} :

$$\mathbf{x} = \mathbf{P} \mathbf{X} \Rightarrow$$

79 CENTRAL PROJECTION FOR PINHOLE

$$\mathbf{P} = \begin{pmatrix} F & 0 & 0 \\ 0 & F & 0 \\ 0 & 0 & 1 \end{pmatrix} \quad (\text{with focal length } f)$$

IN HOMOGENIOUS

$$\mathbf{K} = \begin{pmatrix} F & f & 1 \end{pmatrix}$$

$$\text{Ex: } \begin{pmatrix} F & 0 & 0 \\ 0 & F & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} f_x \\ f_y \\ z \end{pmatrix} \sim \begin{pmatrix} \frac{f_x}{z} \\ \frac{f_y}{z} \\ 1 \end{pmatrix} \approx \begin{pmatrix} f_x \\ f_y \\ f \end{pmatrix} \text{ where } x = \frac{f_x}{z}, y = \frac{f_y}{z}$$

- USING OFFSET

$$\text{camera param } \mathbf{K} = \begin{pmatrix} F & P_x & 0 \\ 0 & F & P_y \\ 0 & 0 & 1 \end{pmatrix}$$

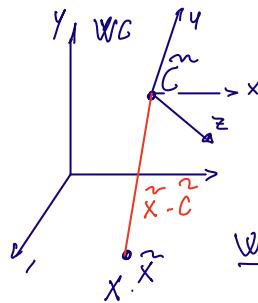
$$\mathbf{P} = (\mathbf{P}_x, \mathbf{P}_y)^T = \text{OFFSET}$$

$$\mathbf{P} = \mathbf{K} \begin{bmatrix} I | 0 \end{bmatrix} = \begin{pmatrix} F & P_x & 0 \\ 0 & F & P_y & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

$$\text{Ex: } \begin{pmatrix} F & P_x & 0 \\ 0 & F & P_y & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} \sim \begin{pmatrix} Fx + P_x z \\ Fy + P_y z \\ z \end{pmatrix} = \begin{pmatrix} \frac{F_x}{z} + P_x \\ \frac{F_y}{z} + P_y \\ 1 \end{pmatrix}_{\text{HOMO}}$$

(80)

NON-CENTRAL ORIENTATION PROJECTION FOR PINHOLE



Express any point from world coord system into camera coordinate system

WHY? So we can do projection into camera image

EXPRESS WORLD POINT IN CAMERA COORDS

$$\tilde{X}$$

world

rotation matrix of camera orient.

world coords of camera center

$$\tilde{X} = \begin{pmatrix} R & -RC \\ O^T & 1 \end{pmatrix} \tilde{x}$$

point coord in camera reference

$$\text{Ex: } \begin{pmatrix} \tilde{X} \\ 1 \end{pmatrix} = \begin{pmatrix} R & -RC \\ O^T & 1 \end{pmatrix} \begin{pmatrix} \tilde{x} \\ 1 \end{pmatrix} = \begin{pmatrix} R\tilde{x} - RC \\ 1 \end{pmatrix}$$

- NO HOMO:

$$\tilde{X} = R\tilde{x} - RC$$

of rotation frame
to world frame

Distance between
world, camera center
and world \tilde{X}

$$R = \text{ortho matrix} = \begin{pmatrix} \hat{x}^T \\ \hat{y}^T \\ \hat{z}^T \end{pmatrix}$$

$\hat{x}, \hat{y}, \hat{z}$ are
unit vectors
of world coords
of camera
axes

reexplained in (81)

(81) GET WORLD COORDS OF POINT FROM
CAMERA COORD

To get from camera coord $\begin{pmatrix} x \\ y \\ z \end{pmatrix}$ to world $\begin{pmatrix} \tilde{x} \\ \tilde{y} \\ \tilde{z} \end{pmatrix}$

$$\tilde{\mathbf{x}} = x \cdot \hat{\mathbf{x}} + y \cdot \hat{\mathbf{y}} + z \cdot \hat{\mathbf{z}} + \tilde{\mathbf{c}}$$

(World coordinate)
 of camera x, y, z
 axis

| Camera Center
 IN WORLD
 COORDINATE

$$\tilde{\mathbf{x}} = (\hat{\mathbf{x}}, \hat{\mathbf{y}}, \hat{\mathbf{z}}) \begin{pmatrix} x \\ y \\ z \end{pmatrix} + \tilde{\mathbf{c}}$$

R⁻¹

$$\tilde{\mathbf{x}} = R^{-1} \mathbf{x} + \tilde{\mathbf{c}} \Leftrightarrow R(\tilde{\mathbf{x}} - \tilde{\mathbf{c}}) = \mathbf{x}$$

$\hat{\mathbf{x}}, \hat{\mathbf{y}}, \hat{\mathbf{z}}$: Unit vectors representing world coords
of camera 3x3s

$$R = \text{orthogonal matrix} = \begin{pmatrix} \hat{\mathbf{x}}^T \\ \hat{\mathbf{y}}^T \\ \hat{\mathbf{z}}^T \end{pmatrix}$$

$$R^{-1} = R^T$$

(82) 60 FROM WORLD TO PIXEL

Rotation matrix: $(\tilde{x}, \tilde{y}, \tilde{z})^T$

$X = \begin{pmatrix} R & -RC \\ O^T & 1 \end{pmatrix} \tilde{X}$

From
World coordinate
to
Camera coordinates
in homogeneous

Point in world coord

TRANSFORM FROM WORLD TO CAMERA

Cameras Coordinates in 3D

Cameras center in world coordinate

- FULL PROJECTION -

WE COMBINE IT WITH THE PROJ.
TO 2D (THE OLD P!)

camera internal parameters

$\underbrace{\text{World} \rightarrow \text{camera}}_{\text{point in pixels}} \rightarrow \text{camera 3D} \rightarrow \text{camera 2D}$

$X = P \tilde{X}$ where $P = KR \begin{bmatrix} I & -\tilde{C} \\ 0 & 1 \end{bmatrix}$

$\begin{bmatrix} \text{point in pixels} \\ \text{point in 3D world} \end{bmatrix}$

$\begin{bmatrix} KR \\ 0 \end{bmatrix} \begin{bmatrix} I & -\tilde{C} \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} KR & -KRC \\ 0 & 1 \end{bmatrix}$

- FROM CAMERA 3D TO CAMERA 2D

$X = P \tilde{X}$ pixels camera 3D point

BUT! $\tilde{X} = \begin{pmatrix} R & -RC \\ O^T & 1 \end{pmatrix} X$ $X \text{ in world}$

So $P = \underline{9 \text{ DOF}}$

$$\begin{aligned} X &= P \begin{pmatrix} R & -RC \\ O^T & 1 \end{pmatrix} \tilde{X} = K \begin{bmatrix} I & 0 \\ 0 & 1 \end{bmatrix} \begin{pmatrix} R & -RC \\ O^T & 1 \end{pmatrix} \tilde{X} \\ &= K \begin{bmatrix} R & -RC \\ 0 & 1 \end{bmatrix} \tilde{X} = K \begin{bmatrix} R & + \\ 0 & 1 \end{bmatrix} \tilde{X} \\ &\quad + = -RC \\ X &= K \underbrace{\begin{bmatrix} I & -C \\ 0 & 1 \end{bmatrix}}_{P_{\text{NEW}}} \tilde{X} \end{aligned}$$

(83) PINHOLE CAMERA WITH NON-SQUARE PIXELS

$$K = \begin{pmatrix} \partial_x & & x_0 \\ \partial_y & & y_0 \\ & & 1 \end{pmatrix}$$

IF pixels are sized
 m_x, m_y in image
 unit distance
 then:

$$\begin{aligned} \partial_x &= f m_x \\ \partial_y &= f m_y \end{aligned} \quad \left. \begin{array}{l} \text{new} \\ \text{focal} \\ \text{lengths} \end{array} \right\}$$

$$\begin{aligned} x_0 &= m_x p_x \\ y_0 &= m_y p_y \end{aligned} \quad \left. \begin{array}{l} \text{new} \\ \text{principal} \\ \text{point} \\ \text{coordinates} \end{array} \right\}$$

So $P = I_3 \xrightarrow{\text{DOE}}$ in (88)

(84) SKEWNESS

$$K = \begin{pmatrix} \partial_x s & x_0 \\ \partial_y s & y_0 \\ 1 \end{pmatrix}$$

$$P = KR[I | -\tilde{C}] = K[R|+] \quad + = -\tilde{R}\tilde{C}$$

$$11 \xrightarrow{\text{DOE}} \text{rank}(P) = 3 \leftarrow \text{finite projective camera}$$

(85) GIVEN ONLY P : CAMERA PROPERTIES

GIVEN $P = [M | P_4]$ where $M \in \mathbb{R}^{3 \times 3}$
 and $P_4 \in \mathbb{R}^3$

- WORLD COORDINATES OF
 CAMERA CENTER
 $C \in \mathbb{R}^4$ such that $PC = \emptyset$
HOMOGENIOUS
COORDINATE

- FINITE CAMERA : M is Non singular?

↳ For finite camera non sing.
 then $C^+ = (\tilde{C}^+, 1)$ M -non singular

↳ For camera at infinity then
 $C^+ = (d^+, 0)$ and $Md = 0$ and
 camera is singular

$$P = [M | P_4], P \cdot \begin{pmatrix} d \\ 0 \end{pmatrix} = \begin{pmatrix} Md \\ 0 \end{pmatrix} \underset{Md=0}{\approx} \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

point of infinity
 in direction d

(86) PROVE C IS CAMERA CENTER
AND IS NULL-SPACE OF P

- Suppose that C is such that $P \cdot C = 0$

- Consider line between C and some point A_{fc}

then $\underline{x(\lambda)} = \lambda A + (1-\lambda)C \quad \lambda \in \mathbb{R}$
any point on line

WE MAP Point in line through A and C

$$P \cdot x(\lambda) = \lambda PA + \underbrace{(1-\lambda)PC}_{\emptyset} = \lambda PA \quad \underline{\underline{\lambda \neq 0}}$$

THIS MEANS THAT THE WHOLE LINE IS PROJECTED
TO SAME IMAGE POINT PA

WHAT THAT MEANS?

These points must be aligned through
the camera center.

Because it works for any line
between C and A then proved that
C is camera center ~~+~~ projection center

⑧7 COMPUTE CAMERA CENTER C SINGULAR

$$M = (P_1, P_2, P_3) \quad C = (x, y, z, w)^T$$

then $\begin{matrix} \text{matrix!} \\ \leftarrow \end{matrix}$ $P \in \mathbb{R}^3$

$$X = \det(P_1, P_3, P_4) \quad Y = -\det(P_1, P_2, P_4)$$

$$Z = \det(P_1, P_2, P_3) \quad W = -\det(P_1, P_2, P_3)$$

⑧8 COMPUTE CAMERA CENTER C NON SINGULAR

$$M = (P_1, P_2, P_3) \quad C = (x, y, z, w)^T$$

$$\tilde{C} = -M^{-1}P_4 \quad C^T = \underline{\left(\tilde{C}^T, 1\right)}$$

$$P \begin{pmatrix} \tilde{C} \\ 1 \end{pmatrix} = \begin{pmatrix} -P_4 + P_4 = \emptyset \\ 0 \end{pmatrix} = \emptyset \quad \text{HOMOGENOUS!}$$

⑧9 RETRIEVE K AND R FROM P

$$P = [M | P_4]$$

$$\downarrow \quad M = KR \quad \text{using RQ decomposition}$$

$$-P_4 = -MC$$

• RQ DECOMP: $\frac{QR = M^T \pi}{\text{COMPUTE QR}}$ where $\pi = \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}$
NORMAL ON THIS

And the result will be : $K = \pi R^T \pi$
 $R = \pi Q^T$

(90) PROVE $K \cdot R = M$

$$K \cdot R = \underbrace{\pi \pi^T}_{I} \underbrace{R^T}_{\substack{| \\ = M}} \underbrace{\pi \cdot Q^T}_{\substack{| \\ = \pi (M^T \pi)^T \\ | \\ = M}} = \pi R^T Q^T = \pi (QR)^T$$

(91) CAMERA AT INFINITY

M IS SINGULAR (HOW IS P ?)

- AFFINE

LAST ROW = $(0, 0, 0, 1)$ MAPS $\infty \rightarrow \infty$

(92) PROVE IF 3 POINTS ARE COLLINEAR

THEN CONIC IS DEGENERATE

$$\left. \begin{array}{l} p^T C p = 0 \\ q^T C q = 0 \\ r^T C r = 0 \end{array} \right\} \text{3 points on conic}$$

since r is collinear with $p \neq q$, then
 NO-HOMO: $r = (1-\lambda)p + \lambda q$ for some $\lambda \in \mathbb{R}$

AFFINE COMBO
BARYCENTRIC

HOMO: exist $\lambda, \mu \in \mathbb{R} / \{0\}$ then $r = \lambda p + \mu q$

Then $r^T C r = (\lambda p + \mu q)^T C (\lambda p + \mu q)$

$$r^T C r = \underbrace{\lambda^2 p^T C p}_{\emptyset} + \underbrace{2\lambda\mu p^T C q}_{\emptyset} + \underbrace{\mu^2 q^T C q}_{\emptyset}$$

$$\Rightarrow p^T C q = q^T C p = 0$$

You can do it for any other point on the line

so we can conclude $r^T C r = 0$

ANY POINT ON LINE IS ONC for any r

CONC



$\downarrow g_2$

CONTINUT.

- we still need to prove is degenerate

↳ We get $\ell = p \times q$ which is the homogeneous coordinate of the line through p and q
 ℓ is the vector orthogonal both to p and to q .

$$C_p \times \ell = C_p \times (p \times q) = p \cdot \underbrace{(C_p)^T q}_{\text{scalar}} - q \cdot \underbrace{(C_p)^T p}_{\text{scalar}}$$

$\left[A \times B \times C = B \cdot \underbrace{[A^T C]}_{\text{scal}} - C \cdot \underbrace{[A^T B]}_{\text{scal}} \right]$

↓ SAME WAY

$$= \underbrace{p p^T C q}_{\emptyset} - \underbrace{q p^T C p}_{\emptyset} = \emptyset$$

$$C_q \times \ell = \emptyset$$

If cross product of 2 vectors is \emptyset

\uparrow
 $C_p \parallel \ell$ and $C_q \parallel \ell$

so

$$C_p = \lambda \ell \quad \& \quad C_q = \mu \ell \quad \text{for some } \lambda, \mu \in \mathbb{R}$$

LET $\underset{\text{POINT}}{s} = \mu p - \lambda q \Rightarrow C_s = C(\mu p - \lambda q) = \mu C_p - \lambda C_q$
 $\underset{\text{we found}}{= \mu \lambda \ell - \lambda \mu \ell = \emptyset}$

some point s

in nullspace of C

and C is degenerate

WITH $s \neq 0$
 because if
 $s = 0 \Rightarrow \mu p = \lambda q$

\downarrow
 so $p \neq q$
 would be
 same point
 because
 HOMO

(93) COMPUTE THE CAMERA MATRIX

$$P = K [R | -R\tilde{C}] \in \mathbb{R}^{3 \times 4}$$

↓
camera
matrix

11 DOF : 5 internal: in K

6 external: rotation on position \tilde{C}

• GIVEN THE CORRESPONDENCES $X_i \leftrightarrow x_i$
FIND P SUCH THAT:

$$x_i = P X_i \quad \text{for all } i$$

- AS P HAS 11 DOF, WE NEED AT LEAST
5 1/2 CORRESPONDENCES.

WE CAN USE MORE AND FIND LEAST-SQUARES

94 DLT (DIRECT LINEAR TRANSFORM)

HOMO COORDINATES!

Instead of looking at $\mathbf{x}_i = P\mathbf{X}_i$, we need $\lambda_i \mathbf{x}_i = P\mathbf{X}_i$ for some $\lambda \neq 0$

$$\Rightarrow \underline{\mathbf{x}_i \times P\mathbf{X}_i = 0}$$

$$P = \begin{pmatrix} P_1 & P_2 & P_3 & P_4 \\ P_5 & P_6 & P_7 & P_8 \\ P_9 & P_{10} & P_{11} & P_{12} \end{pmatrix} - P_1^T - P_2^T - P_3^T \quad P_1 = \begin{pmatrix} P_1 \\ \vdots \\ P_4 \end{pmatrix} \quad P_2 = \begin{pmatrix} P_5 \\ \vdots \\ P_8 \end{pmatrix} \quad P_3 = \begin{pmatrix} P_9 \\ \vdots \\ P_{12} \end{pmatrix}$$

$$P\mathbf{X}_i = \begin{pmatrix} P_1 & P_2 & P_3 & P_4 \\ P_5 & P_6 & P_7 & P_8 \\ P_9 & P_{10} & P_{11} & P_{12} \end{pmatrix} \mathbf{X}_i = \begin{pmatrix} P_1^T \mathbf{X}_i \\ P_2^T \mathbf{X}_i \\ P_3^T \mathbf{X}_i \end{pmatrix}$$

$$\begin{aligned} \mathbf{x}_i \times P\mathbf{X}_i &= \begin{pmatrix} y_i P_3^T \mathbf{X}_i - w_i P_2^T \mathbf{X}_i \\ w_i P_1^T \mathbf{X}_i - x_i P_3^T \mathbf{X}_i \\ x_i P_2^T \mathbf{X}_i - y_i P_1^T \mathbf{X}_i \end{pmatrix} \\ &= \underbrace{\begin{pmatrix} \mathbf{0}^T & -w_i \mathbf{X}_i^T & y_i \mathbf{X}_i^T \\ w_i \mathbf{X}_i^T & \mathbf{0}^T & -x_i \mathbf{X}_i^T \\ -y_i \mathbf{X}_i^T & x_i \mathbf{X}_i^T & \mathbf{0}^T \end{pmatrix}}_{3 \times 12 \text{ row}} \begin{pmatrix} P_1 \\ P_2 \\ P_3 \end{pmatrix} = \overbrace{A_i \cdot P}^{3 \times 12 \text{ col}} \end{aligned}$$

$$A_i = \begin{pmatrix} a_i^T \\ b_i^T \\ c_i^T \end{pmatrix} \quad \text{row 1, 2, 3}$$

$$\begin{aligned} \text{so if we take } x_i \cdot a_i + y_i \cdot b_i &\Rightarrow \begin{pmatrix} w_i y_i \mathbf{X}_i \\ -w_i x_i \mathbf{X}_i \\ x_i y_i \mathbf{X}_i - x_i y_i \mathbf{X}_i \end{pmatrix} = w_i \begin{pmatrix} y_i \mathbf{X}_i \\ -x_i \mathbf{X}_i \\ \mathbf{0} \end{pmatrix} \\ &\Rightarrow \boxed{-C_i \cdot w_i} \end{aligned}$$

SO WE ONLY NEED:
for each point:

$$\left\{ \begin{matrix} 0 & 0 & 0 & 0 & -w_i(x_r) & -w_i(y_r) & -w_i(z_r) & -w_i(t) & y_i(x_r) & y_i(y_r) & y_i(z_r) & y_i(t) \\ w_i(x_r) & w_i(y_r) & w_i(z_r) & w_i(t) & 0 & 0 & 0 & 0 & x_i(x_r) & x_i(y_r) & x_i(z_r) & x_i(t) \end{matrix} \right\}$$

(gs) IMAGE POINT x ASSOCIATED
TO LINE IN WORLD POINTS

$$x(\lambda) = c + \lambda \begin{pmatrix} (KR)^{-1}x \\ 0 \end{pmatrix}$$

- $KR = M$ $M \rightarrow P = [M | P_4]$

- c camera center

- $PC = 0$
- $c^+ = \underbrace{\left(-M^{-1}P_4 \ 1^{-1} \right)^T}_{\text{Homo}}$

WE START:

$$\begin{aligned} P \cdot X(x) &= P \left(c + \lambda \begin{pmatrix} (KR)^{-1} \cdot x \\ 0 \end{pmatrix} \right) \\ &= [M | P_4] \left(\begin{pmatrix} \tilde{c} \\ 1 \end{pmatrix} + \begin{pmatrix} \lambda M^{-1} \cdot x \\ 0 \end{pmatrix} \right) \\ &= \left(M \tilde{c} + P_4 \right) + \left(\lambda x + 0 \right) = \lambda x \underset{\substack{\uparrow \\ \tilde{c} = -M^{-1}P_4}}{\approx} x \end{aligned}$$

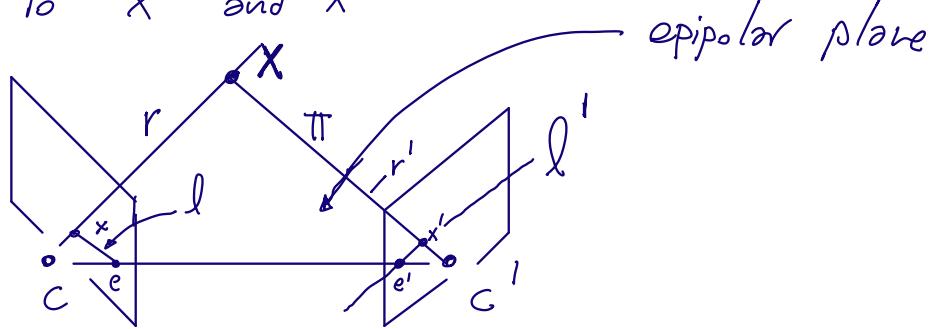
⑨6 STEREO VISION BASICS

USUALLY P maps world points X to image points $x = PX$.

We use correspondences to map $X_i \leftrightarrow x_i$

IF 2 CAMERAS:

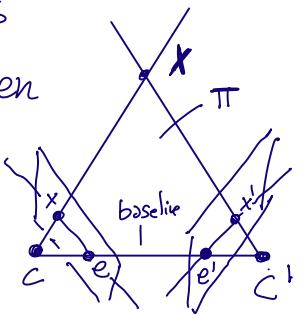
2 Projection matrices P and P' mapping X to x and x'



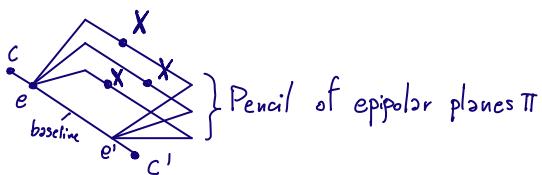
- Given x we know X must lie on ray $r = \vec{Cx}$
- We know x' must lie on the ray r' which is a ray from C' to X $r' = \vec{C'X}$
- So x' must lie on the projection of r on the second image which is $l' =$ epipolar line
- $l' =$ epipolar line is intersection between Π and the second image
- e' is projection of C on second image

(97) EPIPOLAR GEOMETRY

- Π : epipolar plane is uniquely determined by x, C and C'
and always contains the baseline between $C \notin C'$
- $l \notin l'$: epipolar lines are the intersection between the images and Π



VARYING X : As x changes in height, the plane Π will rotate over baseline



(98) FIND ℓ' EPIPOLAR LINE ON 2nd IMAGE

Find mapping from $x \mapsto \ell'$ (USING F)

so: $\ell' = Fx$ F: Fundamental matrix

(98A) PROOF THAT A FUNDAMENTAL MATRIX THAT MAPS X TO ℓ' EXISTS

$\ell' = Fx$ is projection of the ray $r = Cx$ into the second image plane

» We can take 2 points on ray r and project them in 2nd image plane, connect them and have the line.

POINT 1: C , that will project to $e' = P'C$

if P is square, $P^{-1} = P^+$

original camera point

epipole
2nd camera projection matrix
first camera world coords

POINT 2: P^+X We project this point to the 2nd image as: $p'P^+X$

pseudoinverse of P , as P is 3×4 so we can't get inverse, so $P^+ = P^T(PP^T)^{-1}$

$\rightarrow P^+$ will be matrix such that $PP^+ = I \in \mathbb{R}^{3 \times 3}$

HOW TO KNOW THIS IS REALLY A POINT ON RAY $r = Cx$?

Yes!: If we project it in the first image:

$$PP^+X = X \checkmark$$

» So now we connect the 2 points to get ℓ'

$$\ell' = e' \underset{\text{cross prod}}{\underset{|}{\times}} (P'P^+X) = [e']_X \underset{\text{NOTATION IN TOO}}{\underset{|}{\times}} P'P^+X$$

DOESN'T DEPEND ON X ! ONLY ON CAMERAS C, C'

(99) NOTATION OF \times CROSS PRODUCT AS MATRICES

represent $a = \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix}$ AS:

$$[a]_x = \begin{pmatrix} 0 & -a_3 & a_2 \\ a_3 & 0 & -a_1 \\ -a_2 & a_1 & 0 \end{pmatrix}$$

THEN THE CROSS PRODUCT $a \times b$
CAN BE REPRESENTED BY:

$$a \times b = [a]_x b = \underbrace{\left(a^T [b]_x \right)^T}$$

skew
symmetric
matrix

(100) FIND F IF WORLD COORDS
ARE ALIGNED TO FIRST CAMERA
(No ROTATION AND No SHIFT!)

$$P = K \begin{bmatrix} I & 0 \end{bmatrix} = \begin{bmatrix} K & 0 \end{bmatrix}$$

WE CAN ASSUME IT! AS WE CHOOSE THE WORLD

$$P' = K' \begin{bmatrix} R' & t' \end{bmatrix} \quad \text{for second camera}$$

$$\begin{aligned} 1) \text{ LET'S FIND } P^+ &= P^T (P P^T)^{-1} = \begin{bmatrix} K^T \\ 0^T \end{bmatrix} \left(\begin{bmatrix} K & 0 \end{bmatrix} \begin{bmatrix} K^T \\ 0^T \end{bmatrix} \right)^{-1} \\ &= \begin{bmatrix} K^T \\ 0^T \end{bmatrix} (K K^T)^{-1} = \begin{bmatrix} K^{-1} \\ 0^T \end{bmatrix} P^+ \end{aligned}$$

$$2) \text{ LET'S FIND } e' = P' C = \underbrace{\begin{bmatrix} K' R' & K' t' \end{bmatrix}}_{P'} \underbrace{\begin{bmatrix} 0 \\ 1 \end{bmatrix}}_C = \underbrace{K' t'}_{e'}$$

$$\begin{aligned} 3) \text{ LET'S FIND } e = P C &= \underbrace{\begin{bmatrix} K & 0 \end{bmatrix}}_P \underbrace{\begin{bmatrix} \text{WE USE} \\ 0 \end{bmatrix}}_{\text{88}} C \\ &= \underbrace{\begin{bmatrix} K & 0 \end{bmatrix}}_{\substack{H \\ 1}} \underbrace{\begin{bmatrix} - (K' R')^{-1} K' t' \\ P^+ \end{bmatrix}}_{\substack{\text{I} \\ \text{omit } -1}} = K (R')^{-1} \underbrace{\begin{bmatrix} K' \\ I \end{bmatrix}^{-1} K' t'}_I = \underbrace{K (R')^T t'}_e \end{aligned}$$

$$\begin{aligned} \text{88} \\ M' &= K' R' \Rightarrow - (M')^{-1} P^+ = - (K' R')^{-1} K' t' = \boxed{- R'^T t'} \\ P^+ &\approx K' t' \end{aligned}$$

4) Now with e, e' let's get WHY WE CAN? (101)

$$F = [e']_x P' P^T = [e']_x K' R' K^{-1} = (K')^{-T} R'^T K^T [e]_x$$

$$\textcircled{101} \quad [e^r]_x (k' R' k^{-1}) \Rightarrow (k')^{-T} R' k^T [e]_x$$

rule:

$$[w]_x M = M^{-T} [M^{-1} w]_x$$

| |
 matrix non
 vector singular
 matrix

up to scale factor determinant(M) WITH NO-HOMO

- BUT BECAUSE WE HAVE HOMO:

so if $\lambda \neq 0$ scaling doesn't matter

SO IN OUR CASE:

$$\begin{aligned}
 [e^r]_x (k' R' k^{-1}) &= k'^{-T} [k'^{-1} e^r]_x R' k^{-1} = \\
 &\quad \xrightarrow{\text{because } e^r = k'^r \text{ in } \textcircled{100}} \\
 &= k'^{-T} [t^r]_x R' k^{-1} = k'^{-T} R'^{-1} [R'^T t^r]_x k^{-1} = k'^{-T} R' k^T [\underbrace{k R'^T t^r}_e]_x = k'^{-T} R' k^T [e]_x
 \end{aligned}$$

(102) GENERAL CASE OF FINDING F

$$P = \begin{bmatrix} M & P_4 \end{bmatrix} \quad P' = \begin{bmatrix} M' & P'_4 \end{bmatrix}$$

then

$$e = -MM'^{-1}P'_4 + P_4 \quad e' = -M'M^{-1}P_4 + P'_4$$

$$F = [e']_x M' M^{-1} = M'^{-T} M^T [e]_x$$

FIND e and e' (Proof)

$$\begin{aligned} e &= PC' = \begin{bmatrix} M & P_4 \end{bmatrix} \begin{bmatrix} -M'^{-1}P'_4 \\ 1 \end{bmatrix} \xrightarrow{\text{Because } \textcircled{22}} \\ &\quad \left| \begin{array}{c} P \\ C' \end{array} \right. \\ &= -MM'^{-1}P'_4 + P_4 \checkmark \end{aligned}$$

$$e' = P'C = \begin{bmatrix} M' & P'_4 \end{bmatrix} \begin{bmatrix} -M^{-1}P_4 \\ 1 \end{bmatrix} = -M'M^{-1}P_4 + P'_4 \checkmark$$

1) GET POINTS

FIRST POINT: $C \rightarrow e'$

SECOND POINT:

$P_{\infty} = \text{Point on } r \rightarrow P'$

$$P' = \begin{bmatrix} P_{\infty} \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} M' & P'_4 \end{bmatrix} \begin{bmatrix} M^{-1}x \\ 0 \end{bmatrix} = M'M^{-1}x$$

$$r(\lambda) = C + \lambda \begin{pmatrix} M^{-1}x \\ 0 \end{pmatrix} \quad \begin{array}{l} \text{any point on ray} \\ \downarrow \\ \text{Point on ray at } \infty \end{array}$$

2) GET LINE BETWEEN POINTS

$$Q' = e'x(M'M^{-1}x)$$

cross product

$$= \underbrace{[e']_x M' M^{-1} x}_{F} \quad \text{where } F = M'^{-T} M^T [e]_x$$

$\underbrace{[e']_x P' P^+}_{\text{But THIS IS NOT EQUAL } M'M^{-1}!} \quad \text{For some reason...}$

(103) PROPERTIES OF F ($x'^T F x = 0$)

$$x'^T F x = 0 !$$

- $l' = Fx$ (98), x' lies on l' (OBVIOUS)

$$x'^T F x = x'^T l' = 0$$

(104) PROPERTIES OF F

F is unique 3×3 HOMO MATRIX WITH RANK = 2
WITH PROPERTY $x'^T F x = 0$

IT'S RANK 2 BECAUSE $[e^i]_x$ IS RANK DEFICIENT

Because: $[e^i]_x \cdot e^i = e^i \times e^i = 0$

(105) OTHER PROPERTIES OF F:

- F Fundamental (P, P') $\Leftrightarrow F^T$ Fundamental (P', P)
- $l' = Fx$ is epipolar line of x
- $l = F^T x'$ is epipolar line of x'
- F has 7 DOF
- IF $F = [\emptyset]$, both cameras are in the same position

$$\textcircled{106} \quad F e = 0 = e'^T F \quad \text{PROOF}$$

for any $x \neq e$, $l' = Fx$ contains e'

$$\hookrightarrow \underbrace{e'^T F x}_{} = 0 \quad \forall x \neq e$$

when can this be 0? $\Rightarrow e'^T F$ must be \emptyset

$$F^+ e' = \emptyset$$

$\textcircled{107}$ DECOMPOSE F MATRIX

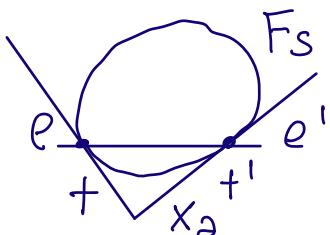
$$F = F_S + F_\partial \text{ where}$$

- F_S is symmetric and represent the STEINER CONIC where the epipoles e, e' lie on.

- F_∂ is skew-symmetric and can be expressed

$$\text{as } F_\partial = [x_\partial]_x$$

where the polar l_∂ of x_∂ intersects F_S at e and e'



e is right null space of F
 e' is left null space of F

(108) PROVE e, e' EPIPOLES LIE ON F_s
THE STEINER CONIC

Since $F_e = \emptyset$ then $\Rightarrow e^T F_e e = \emptyset$

$$\Leftrightarrow e^T (F_s + F_d) e \Rightarrow e^T F_s e + e^T F_d e = \emptyset$$

? $= \emptyset$.

PROVE $e^T F_d e = \emptyset$

\Downarrow

F_d IT'S A SKEW
SYMMETRIC MATRIX

where $F_d = [x_d]_X$

\Downarrow

FROM (99)

$$e^T [x_d]_X e = e^T \underbrace{(x_d \times e)}_{}$$

this is Perpendicular to x_d and e

so $e^T (\text{something } \perp \text{ to } e) = \emptyset$

BUT AS WE HAVE $e^T F_s e + e^T F_d e = \emptyset$

\Downarrow

$$e^T F_s e = \emptyset$$

so e lies on conic F_s !

SAME CAN BE DONE FOR e'

(109) FIND TANGENT TO F_S STEINER CONIC
ON e and e' and X_2 (intersection of them)

1) Find tangent lines by:

$$F_S \cdot e = t \quad F_S \cdot e' = t'$$

2) Find X_2 by:

$$+ x t' \\ \text{cross product}$$

X_2 is the POLE of l_2

l_2 is the POLAR of X_2

(110) PROOF THAT X_2 IS POLE OF l_2

$\because l_2$ IS POLAR OF X_2

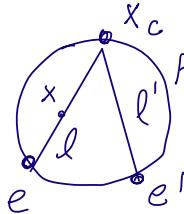
We need to prove that $e \notin e'$ are on $F_S X_2$

We start by showing $e^T F X_2 = 0$ and $e'^T F X_2 = 0$
as given in (108)

We need now to prove that $F_S X_2 = F X_2$

$$\begin{aligned} F_S X_2 &= (F - F_2) X_2 = F X_2 - F_2 X_2 \quad \text{but } F_2 = [X_2]_x \\ &= F X_2 - \underbrace{[X_2]_x}_{0} X_2 \Rightarrow F_S X_2 = F X_2 \checkmark \end{aligned}$$

(11) GEOMETRIC CONSTRUCTION OF EPIPOCAL LINE $\ell' = F_x$



1) Find intersection x_c of l and F_s through e and x

2) Construct line $\ell' = F_x$ through x_c and e'
 ↳ projected
 on second image

1) FIND x_c $\begin{array}{c} \text{vector from } e \text{ to } x \\ \text{PARAMETRIC} \end{array}$

$$l = e + \lambda(x - e)$$

↖ parameter (scale factor)

SOLVE FOR λ : $l^T F_s l = 0 \rightarrow$ get 2 values of λ

We care for x_c !

So we use λ corresponding to x_c

in eq: $l = e + \lambda(x - e)$ to get x_c

2) Now that we have x_c we use it to find ℓ'
 as:

$$x_c \underset{\text{cross product}}{\times} e' = \ell' = F_x$$

112) PROJECTIVE INVARIANCE OF F FUNDAMENTAL MATRIX

$$x \mapsto l^T = Fx \quad \text{and} \quad x^T Fx = 0 \quad \text{are}$$

Projective relationships

>> IF WE TRANSFORM IMAGES WITH
PROJECTIVE TRANSFORMATIONS $H \in \mathbb{R}^{3 \times 3}$

AND $H' \in \mathbb{R}^{3 \times 3}$, THEN THE NEW

FUNDAMENTAL IS $F = H^{I-T} F H^{-1}$

Proof:

$$\begin{aligned} \hat{x} &= Hx \\ \hat{x}' &= H'^{-1}\hat{x}' \end{aligned} \quad \left. \begin{array}{l} \uparrow \\ \downarrow \end{array} \right\} \rightarrow \begin{aligned} \hat{x}'^T F \hat{x} &= x'^T H'^T \underbrace{\left(H'^{-T} F H^{-1} \right)}_{F} \underbrace{\left(H \right)}_{x} = \emptyset \\ &= x'^T H'^T \underbrace{\left(H'^{-T} F H^{-1} \right)}_{I} \underbrace{\left(H \right)}_{I} = \emptyset \\ &= x'^T F x = \emptyset \end{aligned}$$

Proof 2: $\hat{x} = \hat{F}_x$?

$$\hat{L} = H^{1-T} L' = H^{1-T} F_X \stackrel{\text{WE ADD IDENTITY}}{=} H^{1-T} F \underbrace{H^{-1}}_I H_X = \underbrace{H^{1-T} F}_{\neq *} \underbrace{H^{-1}}_{X'} H_X$$

(113) $(P, P') = (PH, P'H)$
 F IS INDEPENDENT FROM WORLD COORD.

Theorem: if $H \in \mathbb{R}^{4 \times 4}$ represents a spatial projective transformation; the fundamental matrixes for pairs (P, P') and $(PH, P'H)$ are identical

Proof:

For any X , the points $x = PX$ and $x' = P'X$ correspond as X is same

$$\text{So: } x^T F x = 0 \Rightarrow \underbrace{X^T P^T}_{x'^T} \underbrace{F}_{F} \underbrace{P X}_{x} = 0$$

Now X with H^{-1} to $\hat{X} = H^{-1}X$

$$\text{then } \hat{x} = (PH)\hat{X} = PHX = x \quad \left. \begin{array}{l} \\ \end{array} \right\} \hat{x}^T F \hat{x} = \\ \underline{\underline{\text{and}}} \quad \hat{x}' = (P'H)\hat{X} = P'HX = x' \quad \left. \begin{array}{l} \\ \end{array} \right\} x'^T F x = 0$$

So F is ALSO FUNDAM FOR $(PH, P'H)$

(114) FIND MATRIX H THAT
GIVES CANONICAL FORM
OF (P, P')

\Rightarrow CANONICAL FORM $\rightarrow P = \underline{[I | 0]}$

Augment P by a 4th row to a non-singular 4x4
matrix $\rightarrow P^* = \begin{bmatrix} P \\ r^T \end{bmatrix}$
 row vector
 linearly indep. from P

Then $H = P^{*-1}$

Lets prove that we get $P^* \cdot H = \begin{bmatrix} I & 0 \\ 0^T & 1 \end{bmatrix}$

$$\begin{aligned} &= \begin{bmatrix} P \\ r^T \end{bmatrix} \cdot H \\ &= \begin{bmatrix} P \cdot H \\ r^T H \end{bmatrix} \rightarrow P \cdot H = [I | 0] \end{aligned}$$

(115) ASSUMING $P' = [M'|P_4']$ AND WE ARE
IN CANONICAL $\Rightarrow F = [P_4']_x M'$

$$P = [M|P_4] \nmid P' = [M'|P_4'] \Rightarrow F = [e']_x M' M^{-1}$$

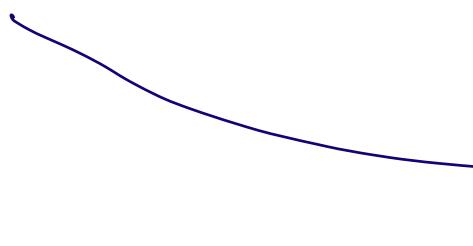


$$e' = -M'M^{-1}P_4 + P_4'$$

$$P = [I|0] \nmid P' = [M'|P_4']$$

AS (402)

$$e' = -M'I \cdot 0 + P_4'$$



$$F = [P_4']_x M'$$

$$F = [P_4 \times P_1 | P_4 \times P_2 | P_4 \times P_3]$$

116 GIVEN F , FIND (P, P') IN CANONICAL FORM

if $F = [\partial]_x A = [\tilde{\partial}]_x \tilde{A} \in \mathbb{R}^{3 \times 3}$ with $\text{rank}(F)=2$

$\begin{matrix} \downarrow \\ \text{fundamental} \\ \text{matrix} \end{matrix}$ $\begin{matrix} \downarrow \\ \text{with} \\ \partial \text{ and } \tilde{\partial} \\ \text{as vectors} \end{matrix}$

\Rightarrow then $\tilde{\partial} = k\partial$ with some $k \neq 0$ but very small!

$$\Rightarrow \tilde{A} = (A + \partial v^+) / k \quad \text{PROVE}$$

Proof of Lemma:

$$\partial^T F = \partial^T [\partial]_x A = \underbrace{[\partial]_x}_F \underbrace{\partial^T}_{\substack{\text{zero} \\ \text{row} \\ \text{vector}}} = 0 \quad \text{because } \partial^T [\partial]_x y \underbrace{|}_{\substack{\text{any} \\ \text{vector}}} \underbrace{\partial^T (\partial_x y)}_{\substack{\text{cross} \\ \text{ortho}}} = 0$$

Likewise $\tilde{\partial}^T F = 0 \Rightarrow$ so ∂ and $\tilde{\partial}$ are on the left null-space of F

but as $\text{rank}(F)=2$ then $\tilde{\partial} = k\partial$ for some $k \neq 0$

$$2) [\partial]_x A = [\tilde{\partial}]_x \tilde{A} = k [\partial]_x \tilde{A} = [\partial]_x k \tilde{A}$$

$$[\partial]_x (k \tilde{A} - A) = 0 \Rightarrow [\partial]_x \left(\underbrace{\text{either of}}_{\substack{\text{of} \\ \text{the 3 columns}}} \right) = 0 \Rightarrow k \tilde{A} - A = [\partial_1 | \partial_2 | \partial_3]$$

THEN $\partial_x \partial_i = 0$ for $i=1,2,3$

$$\partial_i = V_i \cdot \partial \text{ for some } V_i \quad \text{let } V \text{ be } V = \begin{pmatrix} V_1 \\ V_2 \\ V_3 \end{pmatrix} \Rightarrow k \tilde{A} - A = [V_1 \partial | V_2 \partial | V_3 \partial]$$

$$\tilde{A} = (A + \partial v^+) / k \quad \checkmark$$

$\partial \cdot v^T$

(117) IF F IS FUNDAMENTAL MATRIX
OF (P, P') and (\tilde{P}, \tilde{P}') THEN
 $\tilde{P} = PH$ AND $\tilde{P}' = P'H$ for some H

» IF (P, P') and (\tilde{P}, \tilde{P}') are in canonical form

$$\text{ex: } P = [I | 0] = \tilde{P}$$

$$\text{and } P' = [A | \partial] \text{ and } \tilde{P}' = [\tilde{A} | \tilde{\partial}]$$

then F is fundamental mat so:

$$F = [\partial] \times A = [\tilde{\partial}] \times \tilde{A} \text{ and from (116) lemma}$$

$$\tilde{A} = (A + \partial v^T)/k, \quad \tilde{\partial} = k\partial \quad k \neq 0$$

$$\text{Now we let } H = \begin{bmatrix} I/k & 0 \\ v^T/k & k \end{bmatrix}$$

we need to verify that H exists:

$$\bullet PH = [I | 0] \begin{bmatrix} I/k & 0 \\ v^T/k & k \end{bmatrix} = \left[\frac{I}{k} | 0 \right] = \frac{1}{k} \tilde{P} = \tilde{P} \quad \boxed{\text{HONO DISAPPEARS}}$$

$$\bullet P'H = \left[A/k + \partial v^T/k | \partial k \right] = \left[(A + \partial v^T)/k | \partial k \right] \\ = \left[\tilde{A} | \tilde{\partial} \right] = \tilde{P}'$$

(118) GIVEN F AND A SKEW-SYMMETRIC SER^{3x3}

THEN F IS THE FUNDAMENTAL OF

(P, P') FOR $P = [I | 0]$, $P' = [SF | e']$

1. A is skew-symmetric $\Leftrightarrow A^T = -A \Leftrightarrow x^T A x = 0 \forall x$

" \Rightarrow " if $A^T = -A$ then $x^T A x = x^T A^T x = x^T (-A)x$
 we can
 Transpose
 \Downarrow $= -x^T A x$

so IF $x^T A x = -x^T A x$
 THEN $x^T A x = 0$

" \Leftarrow " $0 = (x+y)^T A (x+y) = \underbrace{x^T A x}_0 + \underbrace{x^T A y + y^T A x}_0 + \underbrace{y^T A y}_0$

so:

$$x^T A y + y^T A x = 0 \Rightarrow x^T (-A)y = \underbrace{x^T A^T y}_{\substack{A^T = -A \\ 4 \times 4}} = x^T A^T y$$

2. If F is fundamental matrix of (P, P') $\Leftrightarrow \boxed{P'^T F P}$ is Skew-symmet.

$$\Leftrightarrow \underbrace{x^T P'^T F P x}_{\substack{\text{for all } Xs}} = 0$$

3. Prove $P'^T F P$ is skew symmetric

$$P'^T F P = [SF | e']^T F [I | 0] = \begin{bmatrix} F^T S^T \\ e'^T \end{bmatrix} [F | 0] = \begin{bmatrix} F^T S^T F & 0 \\ e'^T F & 0 \end{bmatrix}$$

because if S is ss then is S^T .

is skew symmetric
 Left nullspace

(119) FUNDAMENTAL MATRIX

PROPRIETIES RECAP

- F depends only on point corresp.
not on P or P'
- We can find F without camera calibration
- If cameras don't change, F can be used on other images from same camera
- Given F , (P, P') can be computed up to 3D
- We can reconstruct 3D scene using $\{X_i \leftrightarrow x_i\}$

(120) ESTIMATE FUNDAMENTAL F

- $x = (x, y, 1)^T \leftrightarrow x' = (x', y', 1)^T$ corresponding points pair

$$F = \begin{pmatrix} f_1 & f_2 & f_3 \\ f_4 & f_5 & f_6 \\ f_7 & f_8 & f_9 \end{pmatrix}$$

THEN

$$x'^T F x = 0 = (x'x, x'y, x'y', x'y, y'y, y', x, y, 1) \cdot f$$

where $f = (f_1, f_2, f_3, f_4, f_5, f_6, f_7, f_8, f_9)^T$

So for n correspondences $\{x'_i \leftrightarrow x_i\}$

we get the HOMO LINEAR SYSTEM:

$A \overset{\text{Unknowns}}{\cancel{f}} = 0$ for F

$$A = \begin{pmatrix} x'_1 x_1 & x'_1 y_1 & x'_1 y'_1 x_1 & y'_1 y_1 & y'_1 y_1 x_1 & y'_1 y_1 y_1 & \dots & \dots & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & \ddots & \ddots \\ x'_n x_n & x'_n y_n & x'_n y'_n x_n & y'_n y_n & y'_n y_n x_n & y'_n y_n y_n & \dots & \dots & \dots \end{pmatrix} \overset{n \times 9}{R}$$

SOLVE
SYSTEM

Solution if $\text{rank } A \geq 8$
unique if $\text{rank } A = 8$

(121) 8-POINT ALGORITHM (START)

$Af = \emptyset$ as in (120)

It has many solutions if $\text{rank}(A) > 8$

If has one solution if $\text{rank}(A) = 8$
with $\|f\| = 1$

If $\text{rank}(A) > 8$ we use least square
solution that minimizes $\|Af\|$ with $\|f\| = 1$

unit
singular
vector
of A

↓ f will then be the last column of v

in the SVD $A = U\Sigma V^T$

$F = \begin{pmatrix} f_1 & f_2 & f_3 \\ f_4 & f_5 & f_6 \\ f_7 & f_8 & f_9 \end{pmatrix}$ COMPUTED
THIS WAY IS
NOT SINGULAR!

This is because if $\text{rank}(F) = 3$ then

$$e'^T F = F e'^T = 0$$

IMPLY $e' = 0$

the only e' epipole will be
 $e' = 0$!

We have to force \tilde{F} to be rank 2, where
 \tilde{F} is the closest fundamental to F which has rank 3

(122) 8-POINT ALGORITHM (FIND \tilde{F})

We have to use \tilde{F} as using non-singular

F gives this condition:

$$\Rightarrow \text{rank}(A) = 3 \text{ then } Az = 0 \text{ and } A^T z = 0 \\ \text{implies that } z = \emptyset$$

This will make it so that there will not be an epipole $e' \neq 0$

$$\text{because } e'^T l' = e'^T F x = \emptyset \quad \text{Hx}$$

$$\Updownarrow \\ e'^T F = F^T e' = \emptyset \quad \text{so } e' = \emptyset!$$

We have to force \tilde{F} to be rank 2, where \tilde{F} is the closest fundamental to F which has rank 3

- CLOSENESS WILL BE CALCULATED WITH FROBENIUS

$$\text{NORM : } \|A\|_F = \sqrt{\sum_{i=1}^m \sum_{j=1}^n |a_{ij}|^2}$$

- To FIND THE BEST APPROXIMATION :

We use the Eck-Var-Mir algo:

$$\underbrace{\text{SVD of } F}_{\text{DECOMPOSE } F} = U \text{diag}(\sigma_1, \sigma_2, \sigma_3) V^T$$

σ_3 zero 3rd element

$$\tilde{F} = U \text{diag}(\sigma_1, \sigma_2, 0) V^T \quad \text{with } \det = 0$$

THIS WILL MINIMIZE $\|F - \tilde{F}\|_F$ subject to
 $\det(\tilde{F}) = 0$

(123) LEAST SQUARE SOLUTION
TO FIND F IN 8-POINTS IF $\text{Rank} > 8$
(OVERDEFINED)

1) START BY $SVD(A) = UDV^T$

V = eigenvectors of $(A^T A)$

2) TAKE LAST COLUMN OF V

This will be your $(f_1 f_2 f_3 f_4 f_5 f_6 f_7 f_8 f_9)^T$
that we use for \tilde{F}

(124) FIND \tilde{F} CLOSEST TO F IN 8 POINT
IF $\text{Rank} = 3$

1) SVD OF F TO FIND $SVD F = U \Sigma V^T$

• U = eigenvectors of $F F^T$

$$\Sigma = \text{diag}(\sigma_1, \sigma_2, \sigma_3) = \begin{bmatrix} \sqrt{\lambda_{\text{eigen}_1}} & 0 & 0 \\ 0 & \sqrt{\lambda_{\text{eigen}_2}} & 0 \\ 0 & 0 & \sqrt{\lambda_{\text{eigen}_3}} \end{bmatrix}$$

where eigen_i are eigenvalues of $F^T F$ or $F F^T$

• V = eigenvectors of $F^T F$

2) From $\text{diag}(\sigma_1, \sigma_2, \sigma_3)$ PUT $\sigma_3 = 0$

and use this new Σ to get

$$\tilde{F} = U \Sigma^{\text{with } \sigma_3=0} V^T$$

⑫ 7-POINT ALGORITHM

We can use only 7 points
so that $\text{rank}(A) = 7$
IN THIS CASE $A\vec{f} = 0$ is 2-DIMENSIONAL SPACE

SO THAT WE GET 2 \vec{f} results : $\vec{f}_1 \notin \vec{f}_2$

• \vec{f}_1 and \vec{f} will be linearly independent vectors
that consider all combinations of F_1 and F_2
that would be the 2 resulting
fundamentals/ matrices

• We let $\underbrace{F_2}_{\substack{\text{General} \\ \text{Fundamental}}} = \lambda F_1 + (1-\lambda) F_2$ for $F_1 \notin F_2$
and solve $\det(F_2) = 0$ to retrieve λ THEN USE λ in
equation $\underbrace{F_2}_{\substack{\text{Fund}}} = \lambda F_1 + (1-\lambda) F$ to get
One or 3 Fundamentals F_a ✓

126

3D RECONSTRUCTION USING F AND CANONICAL CAMERA PAIR (FIND \mathbf{x} USING \mathbf{x}, \mathbf{x}')

1. Estimate F from points correspondences

2. Compute a canonical camera pair

$$P = [I | 0]$$

$$P' = [[e']_x F | e'] \quad \text{with } e' \text{ left null space of } F.$$

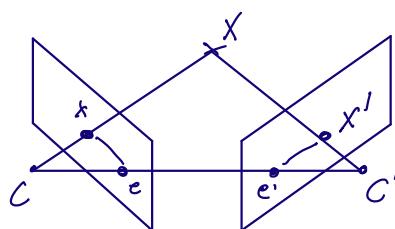
We know that the cameras might need

3. COMMON PROJECTIVE TRANSFORMATION

3. We compute world point \mathbf{x} that projects to the corresponding

$$\mathbf{x} = P\mathbf{x} \text{ and } \mathbf{x}' = P'\mathbf{x} \quad \text{BY TRIANGULATION}$$

TRIANGULATION:



1) intersect rays $r = \overrightarrow{Cx}$ $r' = \overrightarrow{C'x'}$
and get \mathbf{x} ($r \cap r' = \mathbf{x}$)

2) Doesn't work for $x = e \notin x' = e'$

(127) 3D RECONSTRUCTIONS ($P \leftrightarrow \tilde{P}$ USING H)

let $(P, P', \{\mathbf{x}_i\})$ and $(\tilde{P}, \tilde{P}', \{\tilde{\mathbf{x}}_i\})$

[PAGE 266]

be 2 reconstructions from

correspondences $\{\mathbf{x}_i \leftrightarrow \mathbf{x}'_i\}$ then:

$H \in \mathbb{R}^{4 \times 4}$ will be a 3D projective

transformation such that

$$\tilde{P} = PH^{-1}, \quad \tilde{P}' = P'H^{-1} \quad \text{and} \quad \tilde{\mathbf{x}}_i = H\mathbf{x}_i$$

for all points where

$$F\mathbf{x}_i = \mathbf{x}'_i^T F = 0$$

(128) CORRECT PROJECTION

- Know some real world coords \hat{x}_j^*
- Use canonical (P, P') to reconstruct

$$x_j^* \xrightarrow{\text{AS IN } (127)} H \text{ To MAP } \hat{x}_j^* = H x_j^*$$

SO THAT WE HAVE HOMOGRAPHY

THAT RELATES (P, P') TO THE
CORRECT CAMERA PAIR (\hat{P}, \hat{P}')

- H has 15 DOF SO 5 or more \hat{x}_j^*
will suffice