

4-Jan-2019

classmate

Date

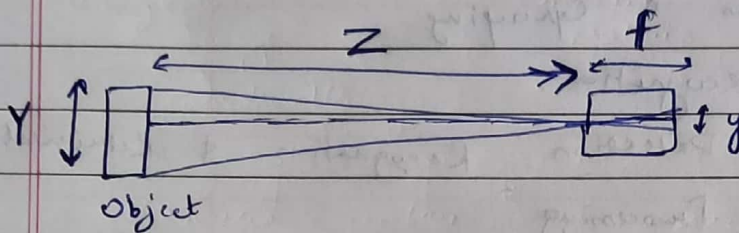
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Imaging and Camera Model

Pinhole Camera

$$y = f \left(\frac{Y}{Z} \right)$$

↓
focal length



⇒ Brightness, sharpness, Exposure Time

VS → PINHOLE

Aperture ↓ DOF ↑
(Depth of field)

$$\text{Focal Ratio} = (f/d)$$

Geometric Distortions

→ Pin cushion, Barrel, Chromatic Aberration

Resolution \Rightarrow Columns - Rows.

Q.

$$\frac{7}{50} = \frac{125}{4}$$

$$\times 2 \quad \frac{50 \times 8.2}{4 \times 7}$$

$$\left(\frac{100}{7}\right) = 19.28 \text{ mm}$$

$$= 1.25 \text{ cm}$$

3000 mm

3000

1

10

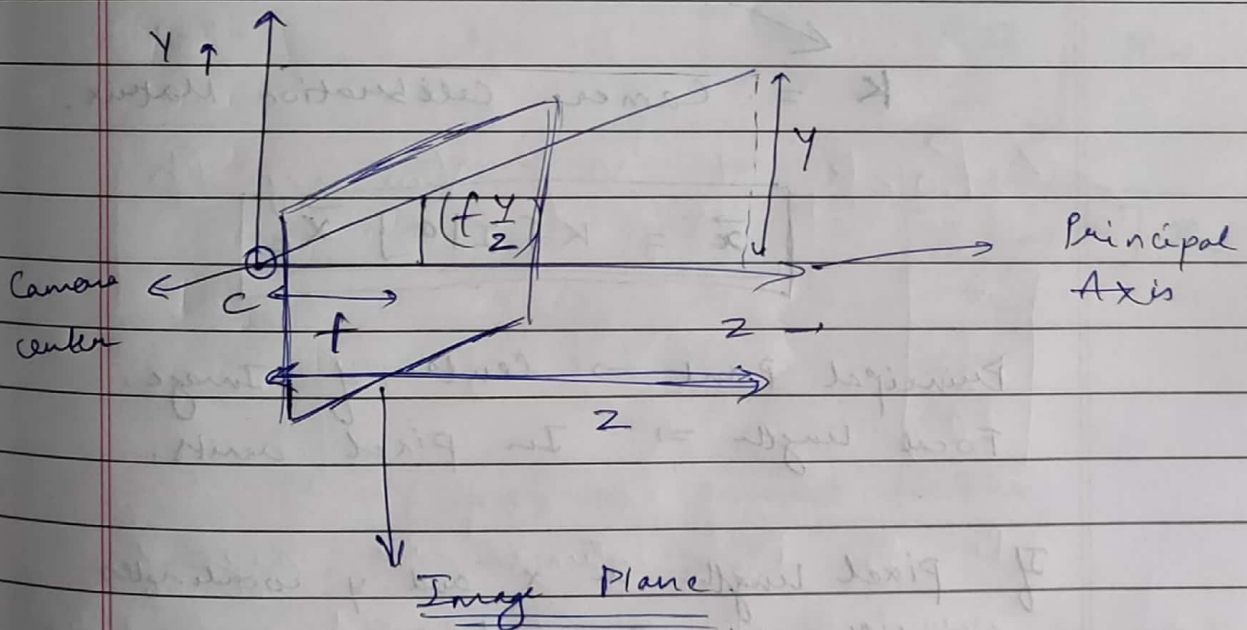
~~428 pixels~~

~~367.5 px~~

1250 Pixels.

Q-1-19.

Perspective Projection



$$x = \frac{f \cdot X}{z}$$

$$y = \frac{f \cdot Y}{z}$$

Non-linear.

HOMOGENEOUS COORDINATES

Convert Non-linear Eq. to linear

$$\begin{array}{ccccc} \vec{x} & & P & & \vec{x}_c \\ \left[\begin{array}{c} x/w \\ y/w \end{array} \right] & \leftarrow & \left[\begin{array}{c} x \\ y \\ w \end{array} \right] & = & \left[\begin{array}{cccc} f & 0 & 0 & 0 \\ 0 & f & 0 & 0 \\ 0 & 0 & 1 & 0 \end{array} \right] \left[\begin{array}{c} x \\ y \\ z \\ 1 \end{array} \right] \\ \downarrow & & \downarrow & & \downarrow \\ \text{Image} & & \text{Homogeneous} & & \text{World} \\ \text{coordinates} & & \text{Image} & & \text{coordinates} \\ & & \text{coordinates.} & & \end{array}$$

Camera Matrix

Camera centre is assumed to be at origin and z-Axis is principal Axis.

$$P = \left[\begin{array}{ccc} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{array} \right] [I | 0]$$

K = Camera Calibration Matrix.

$$\boxed{\vec{x} = K [I | 0] \vec{x}_c}$$

Principal Point \Rightarrow Centre of Image.

Focal length \Rightarrow In pixel units.

If Pixel length in x and y coordinates is different then focal length

+ f_x and f_y exists.

General Camera:

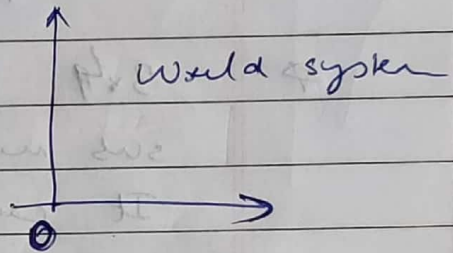
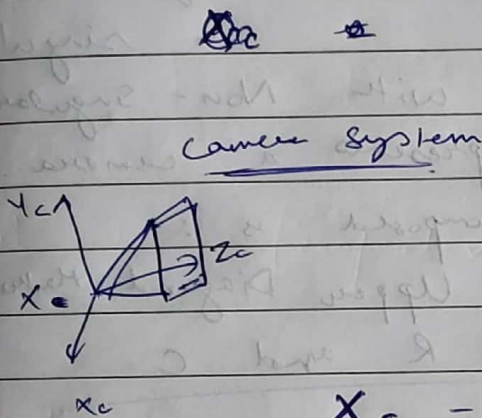
→ Image Center at (x_0, y_0)

Non orthogonal axes with skew s

Upper Diagonal Matrix

$$(K) = \begin{bmatrix} f_x & s & x_0 \\ 0 & f_y & y_0 \\ 0 & 0 & 1 \end{bmatrix}$$

(5 DOF)



$$x_c = \begin{bmatrix} R & -RC \\ 0 & 1 \end{bmatrix} x_w$$

$$x = K [I | 0] x_c = K [R | -RC] x_w$$

↓
P

$P = (3 \times 4) \Rightarrow$ camera Matrix.

$$\rightarrow [P_1 \ P_2 \ P_3 \ P_4]$$

$$\rightarrow \begin{bmatrix} P^1 \\ P^2 \\ P^3 \end{bmatrix}$$

$$\rightarrow [KR | -KRC]$$

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Interest Point Detection

Description

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General camera Egn.

$$x = K [R | -RC] X_w$$

$\underbrace{\hspace{10em}}_{\text{P}}$
 \downarrow
 projects World C to Image - C

* Orthographic projection \Rightarrow Left submatrix is singular.

* 3×4 matrix P with Non-singular left sub matrix represents a camera.

It can be decomposed as :

\rightarrow Non-singular Upper Diagonal Matrix K

\rightarrow Orthogonal vector R and C.

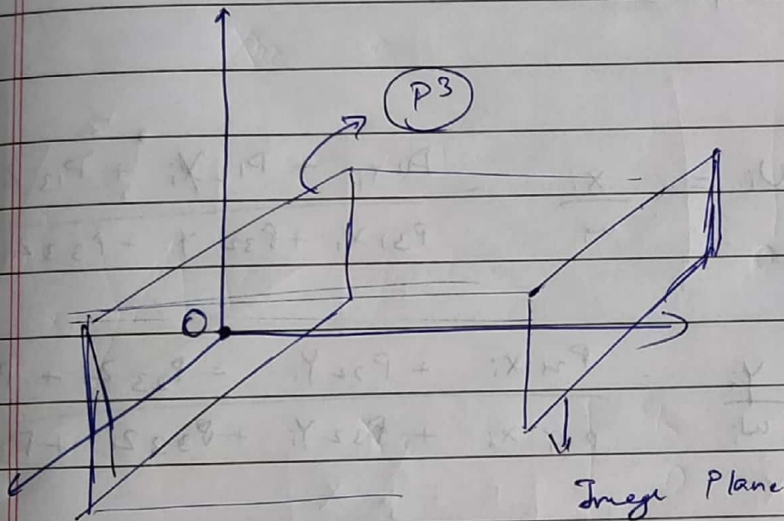
\rightarrow Points at ∞ / Vanishing Pts.

$$\begin{bmatrix} x \\ y \\ z \\ 0 \end{bmatrix}$$

* $\left\{ \begin{array}{l} P_1, P_2, P_3 \text{ columns are images of} \\ \text{vanishing points of world } x, y, z \text{ dir.} \\ P_4 \text{ represents origin coordinates in} \\ \text{Homogeneous frame.} \end{array} \right.$

$$\begin{bmatrix} u \\ v \\ w \end{bmatrix} = \begin{bmatrix} p_{11} & p_{12} & p_{13} & p_{14} \\ p_{21} & p_{22} & p_{23} & p_{24} \\ p_{31} & p_{32} & p_{33} & p_{34} \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ w \end{bmatrix}$$

If $w = 0$ then image pt. lies at ∞ .
 \Rightarrow Row vector P^3 is the principal plane.



Q. Image centre is given by $X_0 = M m_3$
 where m_3 is the third row vector
 of Matrix M .

Q. $\det(M) m_3$ gives the principal axis
 as a vector from camera centre
 through principal pt. to the front
 of camera.

Camera Calibration

- ① 3D Reference Object based calibration
- ② Calibration from precisely Moving plane.
- ③ Calibration using a plane with unknown motion.
- ④ Calibration from set of collinear points that moves s.t. lines passing through fixed pt.

$$u_i = \frac{x_i}{w_i} = \frac{p_{11}x_i + p_{12}y_i + p_{13}z_i + p_{14}}{p_{31}x_i + p_{32}y_i + p_{33}z_i + p_{34}}$$

$$\left\{ \begin{array}{l} v_i = \frac{y_i}{w_i} = \frac{p_{21}x_i + p_{22}y_i + p_{23}z_i + p_{24}}{p_{31}x_i + p_{32}y_i + p_{33}z_i + p_{34}} \end{array} \right.$$

Decomposing P

$$P = K [R | t]$$

$$P = [M | p_4]$$

$$M = KR$$

$$p_4 = Kt$$

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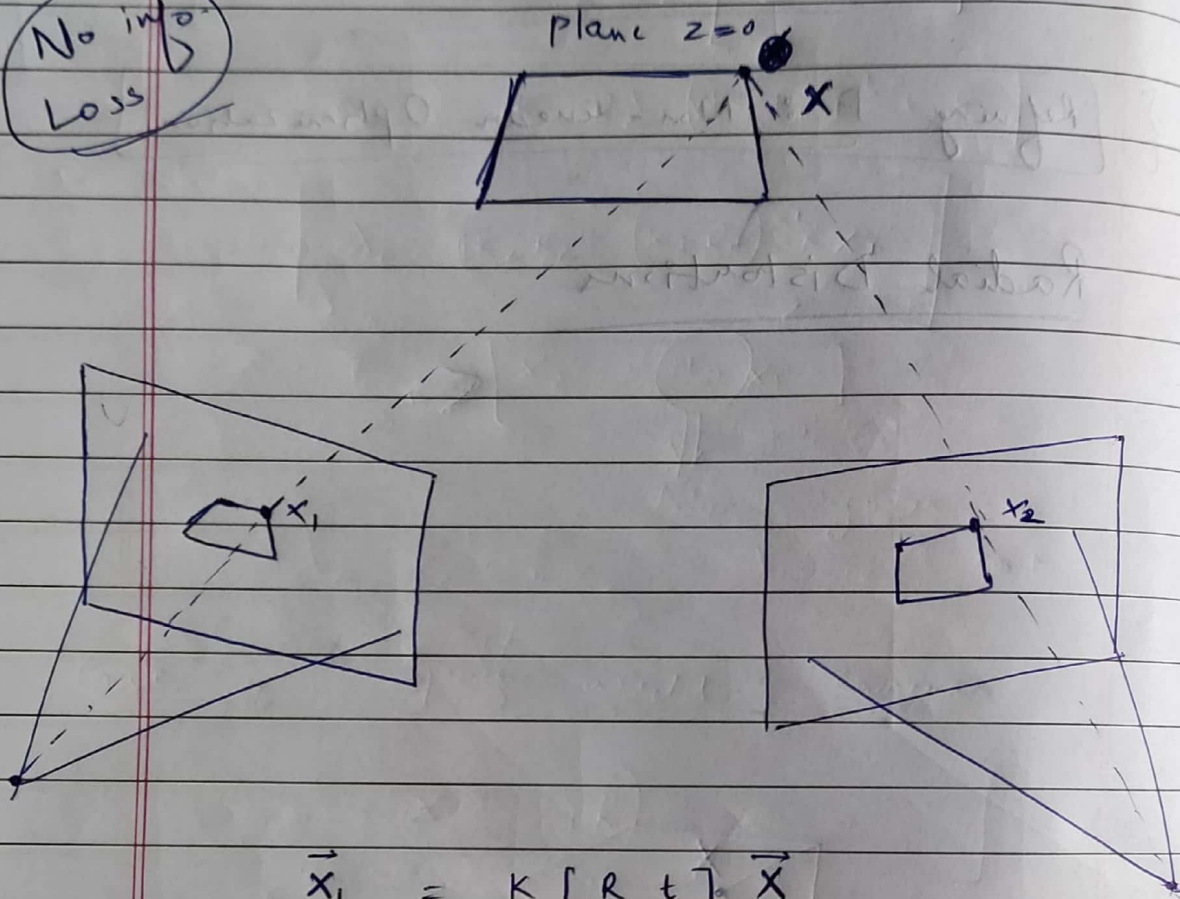
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2 View Geometry

Case-1 Planar World.

No info
Loss



$$\vec{x}_1 = K [R \ t] \vec{X}$$

$$= K \begin{bmatrix} r_1 & r_2 & r_3 & t \end{bmatrix} \begin{bmatrix} x \\ y \\ 0 \\ 1 \end{bmatrix}$$

$$\vec{x}_1 = K \underbrace{\begin{bmatrix} r_1 & r_2 & t \end{bmatrix}}_{(H)} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

$$\vec{x}_1 = H_{3 \times 3} \vec{X}$$

↓
Invertible Matrix

$$\vec{x}_1 = H_1 \vec{x}$$

$$\vec{x}_2 = H_2 \vec{x}$$

Since H is an invertible matrix.

$$\Rightarrow \boxed{x_1 = H_1 H_2^{-1} x_2}$$

$$\boxed{\begin{matrix} \vec{x}_1 = H_{12} \vec{x}_2 \\ \vec{x}_2 = H_{21} \vec{x}_1 \end{matrix}}$$

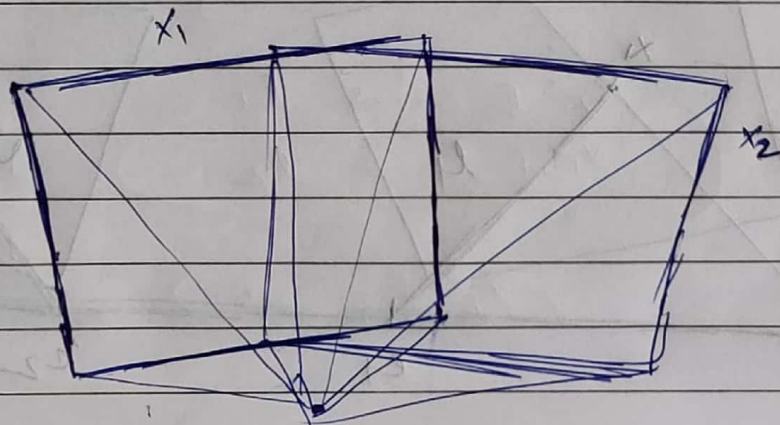
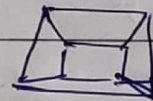
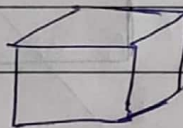
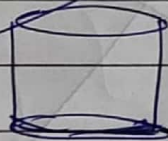
$H \rightarrow$ HOMOGRAPHY MATRIX.

\rightarrow If H is singular then it means that the camera centre lies in the same plane as planar world.

Case-2 Same camera centre.

Arbitrary world

Same info
loss in I_1 and I_2



$$x_1 = K_1 R_1 \begin{bmatrix} I & -c \end{bmatrix} x$$

$$x_2 = K_2 R_2 \begin{bmatrix} I & -c \end{bmatrix} x$$

$$x_2 = K_2 R_2 (K_1 R_1)^T K_1 R_1 \begin{bmatrix} I & -c \end{bmatrix} x$$

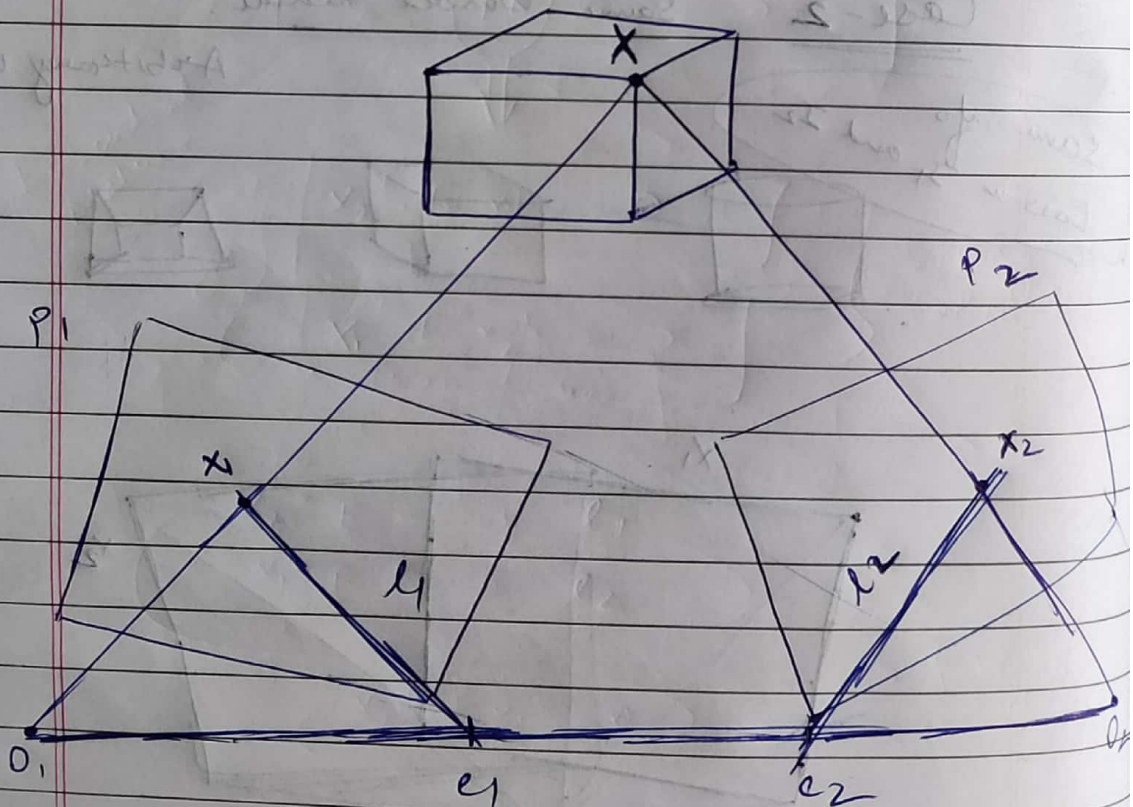
$$x_2 = (K_2 R_2) (K_1 R_1)^{-1} x_1$$

$$\Rightarrow x_2 = H_{21} x_1$$

Non-Singular 3×3 Matrix.

Case - 3

Generic world & cameras



$X, O_1, O_2 \longrightarrow$ Plane

e_1 lies on P_1

e_2 lies on P_2

\Rightarrow Any pt on L_1 lies on line L_2

e_1 is image of O_2 on plane from O_1

$\Rightarrow e_1$ will be a constant

Similarly e_2 is also constant

If pt X is varied then the plane XO_1O_2 ~~varies~~ varies, but e_1 and e_2 does not change.

All world pts that map to x_1 in I_1 (pre-image of x_1) map to line L_2 in I_2 is called epipolar line.

Epipoles

Epipolar Plane

Epipolar Constraints

Considering $K_1, K_2 = I$.

Assuming origin to be at 0,

$$\lambda_1 x_1 = X$$

$$\lambda_2 x_2 = RX + T$$

$$\lambda_2 x_2 = R \lambda_1 x_1 + T$$

$$\lambda_2 \hat{T} x_2 = \lambda_1 \hat{T} R x_1$$

\hat{T} → matrix formulated as $\hat{T} x T$
→ cross product matrix

$$\lambda_2 \bar{x}_2^T \hat{T} \bar{x}_2 = \lambda_1 \bar{x}_2^T \hat{T} R \bar{x}_1$$

↓
0

$$\bar{x}_2^T (\hat{T} R) \bar{x}_1 = 0$$

$E \rightarrow$ Essential Matrix

$$\begin{cases} \bar{x}_2^T E \bar{x}_1 = 0 \\ \bar{x}_1^T E \bar{x}_2 = 0 \end{cases}$$

Strong Calibration

$$\begin{cases} \bar{x}_2^T E \rightarrow l_1 \\ \bar{x}_1^T E \rightarrow l_2 \end{cases}$$

$$\bar{x}_2^T F \bar{x}_1 = 0$$

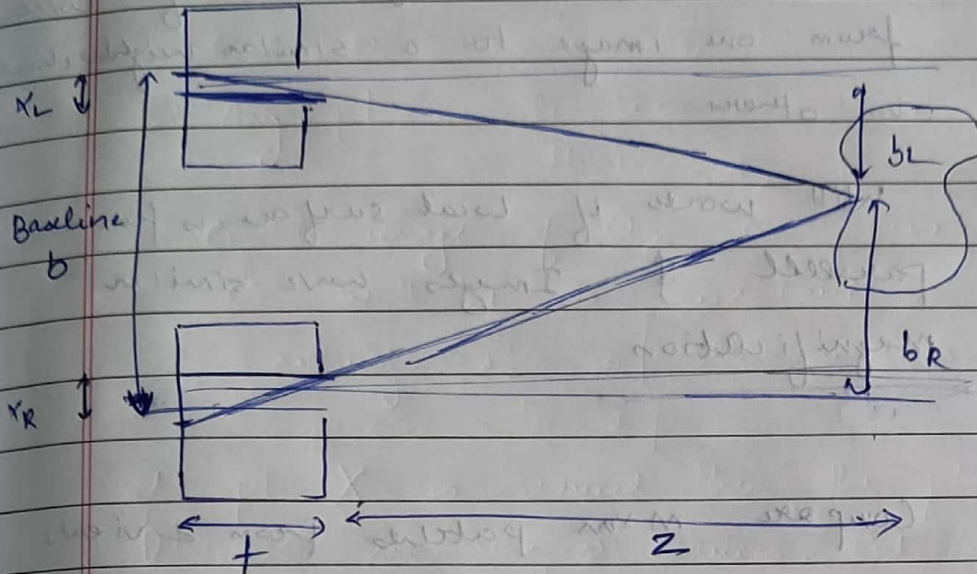
Weak Calibration

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Stereo



$$\frac{b_R + b_L}{Z} = \frac{x_R}{f} + \frac{x_L}{f}$$

$$\left(\frac{b}{Z} \right) = \frac{x_L + x_R}{f}$$

$$Z = \left(\frac{b \cdot f}{x_L + x_R} \right)$$

A Larger baseline can give more reliable depth, however matching becomes harder.

Matching

Match a (small) neighborhood of from one image to a similar neighborhood in others.

Will work if local surface is fronto-parallel & Images have similar magnification.

Compare $m \times m$ patches from 2 views.

Form vector v & v' .

Scores \Rightarrow Sum of Absolute Diff $= \|v - v'\|_1$

Sum of squared dist $= \|v - v'\|_2$

Correlation $\Rightarrow \frac{v'^T v}{\sqrt{v'^T v} \sqrt{v^T v}}$

Normalized Correlation $\Rightarrow \frac{\bar{v}'^T \bar{v}}{\sqrt{\bar{v}'^T \bar{v}} \sqrt{\bar{v}^T \bar{v}}}$

Range $[-1, 1]$

Census Transform

\rightarrow speed

Use Epipolar constraints

Handwritten notes on lined paper:

Case 177

Handwritten text, possibly a signature or name, is visible below the case number.

- If pt A is to left of B in view 1 it will be to the left of B in view 2 also. Eg: DP.

Sparse correspondence \Rightarrow Good feature pt.
 Dense "

- Can be represented using aligning one image to another using H .
- Reproject images onto common parallel plane.

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SIFT INTEREST POINT DETECTOR.

LOG vs DOG

$$DOG(x, y) = \frac{1}{2\pi\sigma_1} e^{-\frac{(x_1^2 + y_1^2)}{2\sigma_1^2}} -$$

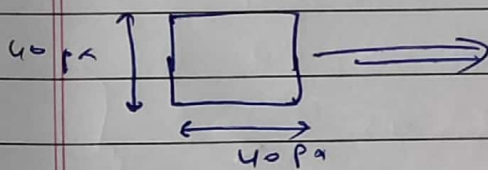
$$\frac{1}{2\pi\sigma_2} e^{-\frac{(x_1^2 + y_1^2)}{2\sigma_2^2}}$$

MOPS Feature Descriptor

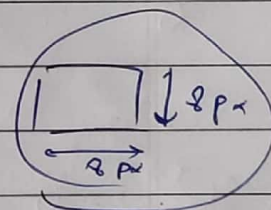
Description is based on nearby image region

- Intensity values
- Moments
- Derivatives
- MOPS / SIFT

MOPS ⇒ 64 length.



subsampled
at 5x



Zero Mean &
Unit Variance