

Camera Calibration :



- How to recover 3D structure of the scene from its images?
- To go from images and full metric reconstruction we need two things →
 - 1) first is position and orientation of camera with respect to the world and coordinate frame. → These are external parameters
 - 2) internal parameters such as focal length
How camera maps their perspective projection on its image plane
- Method to find camera's internal and external parameters
- 1) Linear Camera Model.
turns out to be a single matrix called projection matrix.
- From Rich known geometry of object at various stages camera calibrate using it
To determine the projection matrix intrinsic and extrinsic matrix path

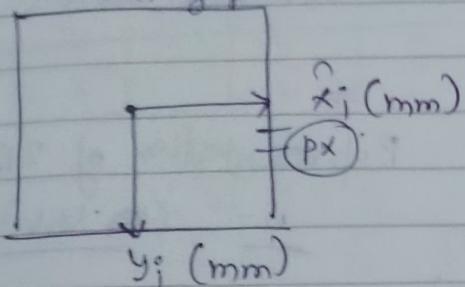
perspective projection

equation

$$\frac{x_i}{f} = \frac{x_c}{z_c}$$

$$\frac{y_i}{f} = \frac{y_c}{z_c}$$

Image plane



$x_i \rightarrow$ projection of point P
 $y_i \rightarrow$ onto image plane

p_x could be rectangular, it depends on px density.

$m_x \rightarrow$ px density in x

$m_y \rightarrow$ px density in y dir

$$u = m_x x_i \\ = m_x \frac{x_c f}{z_c}$$

$$v = m_y y_i = m_y \left(\frac{y_c}{z_c} \right) f$$

we don't know the point of image piercing the plane, usually it is taken at the corner of the plane

that's modify our equation.

$$u = m_x f \frac{x_c}{z_c} + o_x$$

$$v = m_y f \left(\frac{y_c}{z_c} \right) + o_y$$

combining

$$u = f_x \left(\frac{x_c}{z_c} \right) + o_x$$

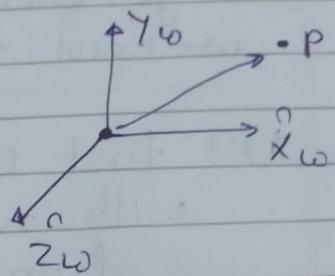
$$v = f_y \left(\frac{y_c}{z_c} \right) + o_y$$

effective focal length

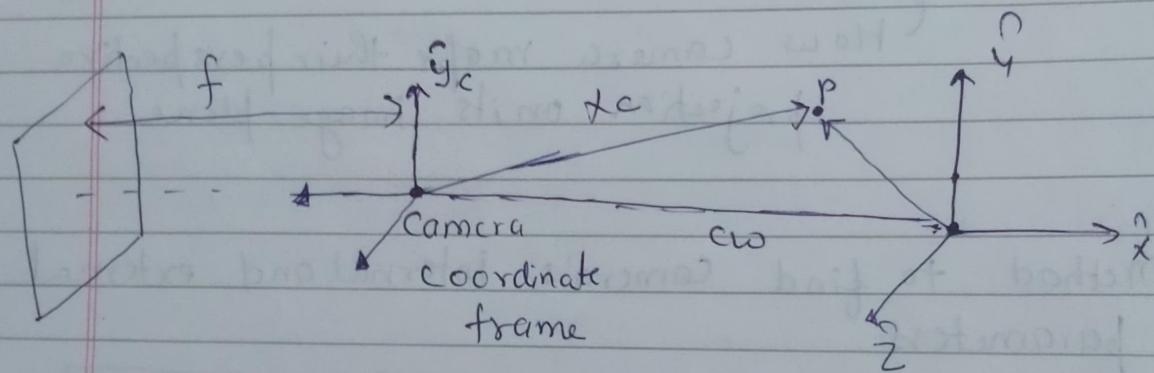
Simple stereo

How to recover 3D structure
from 2 calibrated cameras.

→ Forward Imaging Model
it takes us from 3D to 2D.



world coordinate frame



$z_c \rightarrow$ z_{axis} of CCF is aligned with optical Axis

अगर हम EN 3D में CCF का wrt world c-f mapping find करते हैं तो create an projection

$$x_i = \begin{bmatrix} x_i \\ y_i \end{bmatrix} \quad \xrightarrow{\quad} \quad x_c = \begin{bmatrix} x_c \\ y_c \\ z_c \end{bmatrix}$$

$$x_w = \begin{bmatrix} x_w \\ y_w \\ z_w \end{bmatrix}$$

if you have $x_c \rightarrow$ use $\frac{1}{z}$

perspective projection

perspective projection.

$$u = fx \left(\frac{x_c}{z_c} \right) + o_x \quad (1) \quad v = fy \left(\frac{y_c}{z_c} \right) + o_y$$

$$\begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} \tilde{u} \\ \tilde{v} \\ \tilde{w} \end{bmatrix} = \begin{bmatrix} z_c u \\ z_c v \\ z_c \end{bmatrix} \quad (2)$$

where

$$\begin{aligned} z_c u &= \tilde{u} \\ z_c &= \frac{\tilde{u}}{u} \end{aligned}$$

$$\therefore \begin{aligned} \tilde{w} u &= \tilde{u} \\ u &= \frac{\tilde{u}}{\tilde{w}} \end{aligned} \quad \begin{aligned} \tilde{w} v &= \tilde{v} \\ v &= \frac{\tilde{v}}{\tilde{w}} \end{aligned}$$

on multiplying 1 by z_c

$$z_c \cdot u = fx \cdot x_c + o_x z_c =$$

$$\begin{bmatrix} fx x_c + o_x z_c \\ fy y_c + z_c o_y \\ z_c \end{bmatrix}$$

simply karo

$$\underbrace{\begin{bmatrix} fx & 0 & o_x & 0 \\ 0 & fy & o_y & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}}_{\text{matrix}} \begin{bmatrix} x_c \\ y_c \\ z_c \\ 1 \end{bmatrix}$$

is sinko simply chon
karo

it include all the internal parameter

This is linear model for perspective projection

$(f_x f_y o_x o_y) \rightarrow$ Intrinsic parameters of the camera.

They represent the camera's internal geometry.

Go to linear model | it makes our life easier.

$$\frac{u}{w} \neq 0$$

↳ Homogeneous Coordinates

Representation of 2D point
 $\underline{u} = (\underline{\underline{u}}, v, w)$

is 3D point $\underline{\underline{u}} = (\tilde{u}, \tilde{v}, \tilde{w})$ - w is fictitious

$$u = \frac{\tilde{u}}{\tilde{w}}$$

$$v = \frac{\tilde{v}}{\tilde{w}}$$

$$u = (wv, v, 1)$$

$$\underline{\underline{u}} = u^1, v^1, w^1$$

used for scaling and normalization

$$u = \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} \tilde{u} \\ \tilde{v} \\ \tilde{w} \end{bmatrix} = \begin{bmatrix} \tilde{u} \\ \tilde{v} \\ \tilde{w} \end{bmatrix} = \underline{\underline{u}}$$

$$u = \frac{u^1}{w^1}$$

$$v = \frac{v^1}{w^1}$$

Some you can do with $3D \rightarrow 4D$ point

Rational matrix is orthonormal \rightarrow

$$u^T v = 0$$

in case of
matrix

$$u \cdot v = 0$$

$$\text{dot}(u, v) = 0$$

$$u^T u = 1$$

$$v^T v = 1$$

unit length

eg) - x axis, y axis, z axis

In orthonormal matrix \rightarrow square matrix whose row / col vectors are orthonormal.

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

$$\checkmark R^T R = R R^T = I$$

condition

of normality

$$\rightarrow x_c = R(x_w - c_w)$$

()
 R.m \rightarrow wrt
 C-CF wcf
 = = =

find x_c in world coordinate

$$x_c = (x_w - c_w) R$$

we have mapped to wcf to ccf

$$= x_c = RX_w - RC_w = RX_w + t$$

$$t = -RC_w$$

translation
matrix

$$x_c = \begin{bmatrix} x_c \\ y_c \\ z_c \end{bmatrix} = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} \begin{bmatrix} x_w \\ y_w \\ z_w \end{bmatrix} + \begin{bmatrix} t_x \\ t_y \\ t_z \end{bmatrix}$$

2nd doubt

wcf \rightarrow world coordinate frame

Intrinsic Matrix

$$\begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} \bar{u} \\ \bar{v} \\ \bar{w} \end{bmatrix} = \begin{bmatrix} f_x & 0 & o_x & 0 \\ 0 & f_y & o_y & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_c \\ y_c \\ z_c \\ 1 \end{bmatrix}$$

$$K = \begin{bmatrix} f_x & 0 & o_x \\ 0 & f_y & o_y \\ 0 & 0 & 1 \end{bmatrix} \rightarrow \text{calibration Matrix}$$

it is upper Right Triangular matrix \rightarrow invertible

$M_{int} = [K | 0] \rightarrow$ concatenation of $K | 0$

$$\begin{bmatrix} f_x & 0 & o_x & 0 \\ 0 & f_y & o_y & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \leftarrow$$

mapping the world coordinates to camera coordinates

3D to 3D \rightarrow The position and orientation to

camera coordinate frame to

world coordinate frame

position

c_w

$\rightarrow R$ orientation

3×3 matrix

$$R = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} \rightarrow \begin{array}{l} \text{Dirn of } x_c \text{ in wcf} \\ \text{Dirn of } y_c \text{ in wcf} \\ \text{Dirn of } z_c \text{ in wcf} \end{array}$$

$$\begin{bmatrix} x_c \\ y_c \\ z_c \\ 1 \end{bmatrix} = \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_x \\ r_{21} & r_{22} & r_{23} & t_y \\ r_{31} & r_{32} & r_{33} & t_z \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_w \\ y_w \\ z_w \\ 1 \end{bmatrix}$$

Now solve for each \hat{x}_c

$$M_{ext} = \begin{bmatrix} R_{3 \times 3} & t \\ 0_{1 \times 3} & 1 \end{bmatrix} = \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_x \\ r_{21} & r_{22} & r_{23} & t_y \\ r_{31} & r_{32} & r_{33} & t_z \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\hat{x}_c = M_{ext} \hat{x}_w$$

projection matrix P

$$P = \begin{bmatrix} f_x & 0 & o_x & 0 \\ 0 & f_y & o_y & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_c \\ y_c \\ z_c \\ 1 \end{bmatrix}$$

$$\hat{u} = M_{int} \hat{x}_c$$

$$\begin{bmatrix} x_c \\ y_c \\ z_c \\ 1 \end{bmatrix} = \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_x \\ r_{21} & r_{22} & r_{23} & t_y \\ r_{31} & r_{32} & r_{33} & t_z \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_w \\ y_w \\ z_w \\ 1 \end{bmatrix}$$

$$\hat{x}_c = M_{ext} \hat{x}_w$$

set of corresponding point $\bar{u}_1 \bar{u}_2 \bar{u}_3 \bar{u}_4 \rightarrow$ map $\bar{u}_1 \bar{u}_2 \bar{u}_3 \bar{u}_4$

$$\boxed{\text{Step 3}} \rightarrow \begin{bmatrix} u_i \\ v_i \end{bmatrix}_{3 \times 2} = \begin{bmatrix} p_{11} & p_{12} & p_{13} & p_{14} \\ p_{21} & p_{22} & p_{23} & p_{24} \\ p_{31} & p_{32} & p_{33} & p_{34} \end{bmatrix}_{3 \times 4} \begin{bmatrix} x_w^i \\ y_w^i \\ z_w^i \end{bmatrix}_{4 \times 1}$$

known unknown known

$$= u_i = \frac{p_{11}x_w^i + p_{12}y_w^i + p_{13}z_w^i + p_{14}}{p_{31}x_w^i + p_{32}y_w^i + p_{33}z_w^i + p_{34}}$$

$$v_i = \frac{p_{21}x_w^i + p_{22}y_w^i + p_{23}z_w^i + p_{24}}{p_{31}x_w^i + p_{32}y_w^i + p_{33}z_w^i + p_{34}}$$

$$A \begin{bmatrix} 0 \\ p_{11} \\ p_{12} \\ \vdots \\ p_{34} \end{bmatrix} = 0$$

$AP=0$) equation

for screenshot
at 10:21 AM

Scaling $\|p\|_2 = 1$ impact $\|u\|_2 = 1$

Combining →

$$\tilde{u} = M_{int}M_{ext} \tilde{x}_w = P \tilde{x}_w$$

$\tilde{u} = M_{int}M_{ext}$

points in 3D

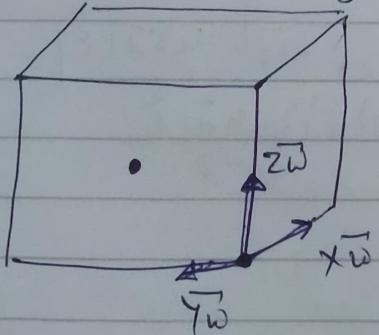
$$\begin{bmatrix} u \\ v \\ w \end{bmatrix} = \begin{bmatrix} p_{11} & p_{12} & p_{13} & p_{14} \\ p_{21} & p_{22} & p_{23} & p_{24} \\ p_{31} & p_{32} & p_{33} & p_{34} \end{bmatrix} \begin{bmatrix} x_w \\ y_w \\ z_w \end{bmatrix}$$

We only need to find projection matrix to calibrate camera

Lecture 3

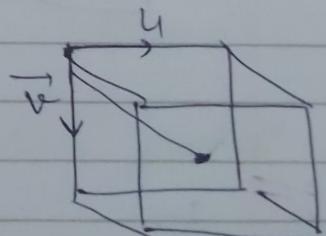
estimation of projection matrix.

1) capture an image of an object with known geometry.



place wcf in one of the corner of the cube

image \rightarrow $\begin{bmatrix} u \\ v \end{bmatrix}$



wcf

$$x_w = \begin{bmatrix} x_w \\ y_w \\ z_w \end{bmatrix} = \begin{bmatrix} 0 \\ 3 \\ 4 \end{bmatrix} \text{ inches}$$

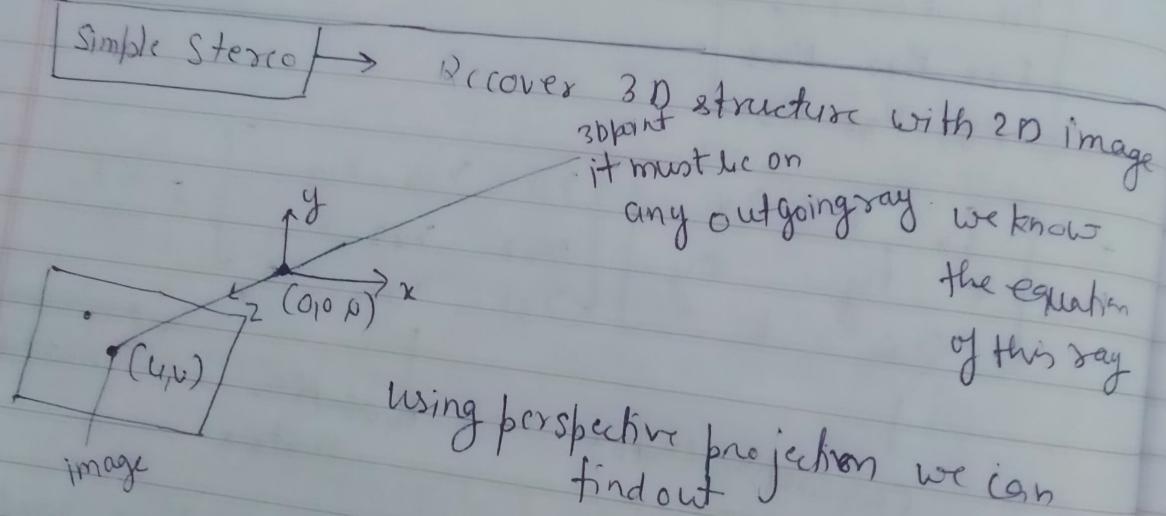
$$\cdot u = \begin{bmatrix} 4 \\ v \end{bmatrix} = \begin{bmatrix} 56 \\ 115 \end{bmatrix} \text{ px}$$

$$\begin{bmatrix} p_{14} \\ p_{24} \\ p_{34} \end{bmatrix} = \underbrace{\begin{bmatrix} f_x & 0 & \alpha \\ 0 & f_y & \beta \\ 0 & 0 & 1 \end{bmatrix}}_K \underbrace{\begin{bmatrix} t_x \\ t_y \\ t_z \end{bmatrix}}_t.$$

$$t = K^{-1} \begin{bmatrix} p_{14} \\ p_{24} \\ p_{34} \end{bmatrix}, \text{ translation matrix}$$

- lenses have distortion
 - ↳ Radial
 - ↳ Tangential
- Pinhole st. distortion
and etc etc

it makes the model non-linear



$$u = f_x \frac{x_c}{z_c} + o_x \quad v = f_y \frac{y_c}{z_c} + o_y$$

2D to 3D

$x_c = ?$

$$\frac{(u - o_x) z_c}{f_x} = x$$

$\boxed{z > 0}$

y_{eff}
Frontal

by rearranging this

Intrinsic and extrinsic matrix

projection matrix can be decomposed further into intrinsic / extrinsic matrix

$$P = \begin{bmatrix} P_{11} & P_{12} & P_{13} & P_{14} \\ P_{21} & P_{22} & P_{23} & P_{24} \\ P_{31} & P_{32} & P_{33} & P_{34} \end{bmatrix} = \begin{bmatrix} f_x & 0 & o_x & 0 \\ 0 & f_y & o_y & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} M_{int}$$

Consider only this

$$\begin{bmatrix} P_{11} & P_{12} & P_{13} \\ P_{21} & P_{22} & P_{23} \\ P_{31} & P_{32} & P_{33} \end{bmatrix}$$

$$\begin{bmatrix} r_{11} & r_{12} & r_{13} & t_x \\ r_{21} & r_{22} & r_{23} & t_y \\ r_{31} & r_{32} & r_{33} & t_z \\ 0 & 0 & 0 & 1 \end{bmatrix} M_{ext}$$

$$= \begin{bmatrix} f_x & 0 & o_x \\ 0 & f_y & o_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} = K R$$

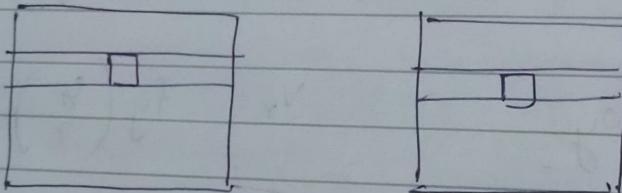
calibration matrix | rotation matrix ↴ orthorhombic matrix

it is possible to find K and R from their product using "QR factorization".

find translational now.

$$P = \begin{bmatrix} - & - & - & p_{14} \\ - & - & - & p_{24} \\ - & - & - & p_{34} \end{bmatrix} = \begin{array}{l} \text{transform, } \\ \text{see next page} \end{array}$$

- Disparity & baseline as per the formula
 - Larger the baseline, more precisely the disparity is calculated.
 - Stereo matching
 - Find the Disparity between left and right stereo pair.
- there is no disparity in Y dirⁿ in case of Horizontal disparity
- $$v_L = v_R = fy \left(\frac{y}{z} \right) + oy$$
- Each corresponding.

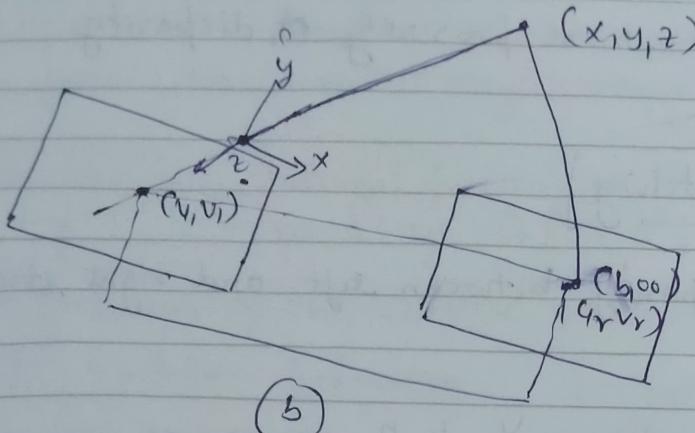


By this we know that this portion must lie on the same horizontal ~~area~~ window.

Issues:

- Texture $\{T\} - \{R\}$
- Should not be overlapping
- Window size $\{W\} - \{R\}$

Wanted more info to construct 3D structure
image out of image



This system is known
as simple stereo
binocular vision

(b) We have 4 eqn

$$u_l = f_x \frac{x}{z} + o_x$$

$$u_R = f_x \left(\frac{x-b}{z} \right) + o_x$$

$$v_l = f_y \frac{y}{z} + o_y$$

$$v_r = f_y \left(\frac{y}{z} \right) + o_y$$

Solving for (x, y, z)

$$x = \frac{b(u_l - o_x)}{u_l - u_r}$$

$$y = \frac{b f_x (v_l - o_y)}{f_y (u_l - u_r)}$$

$$z = \frac{b f_x}{u_l - u_r}$$

Imp
depth

$u_l - u_r$ → Difference in the u coordinate . (Disparity)

disparity is inversely proportional to z

Uncalibrated camera stereo →

Method to estimate 3D structure of a static scene from two arbitrary views.

assume → internal parameters are known to us
 meta tag में दीता है।

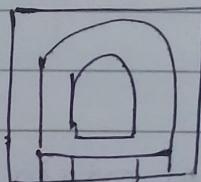
1) problem of uncalibrated stereo.

1) Epipolar geometry | Estimating fundamental matrix

→ finding dense correspondences.

→ computing Depth

1) Problem of uncalibrated Stereo.



Left camera

Right camera
of image

Intrinsics are known to us → meta tag में दीता है।
 $(f_x, f_y, \alpha_x, \alpha_y)$

Extrinsics (relative position | orientation of cameras) unknown.

$$\begin{bmatrix} x_L & y_L & z_L \end{bmatrix} \begin{bmatrix} ty z_L - t_z y_L \\ t_z x_L - t_x z_L \\ tx y_L - ty x_L \end{bmatrix} = 0 \quad \text{cross-product definition}$$

↓

matrix is get converted to Eqn 1

$$\begin{bmatrix} x_L & y_L & z_L \end{bmatrix} \begin{bmatrix} 0 & -t_z & ty \\ t_z & 0 & -tx \\ -tx & tx & 0 \end{bmatrix} \begin{bmatrix} x_L \\ y_L \\ z_L \end{bmatrix} = 0 \quad \text{--- (1)}$$

 $T_X \rightarrow \text{Translation matrix}$

$t_{3xL} \rightarrow$ position of Right camera in Left Camera's frame
 R_{3x3} " Orientation "

$$\boxed{x_L = R x_r + t}$$

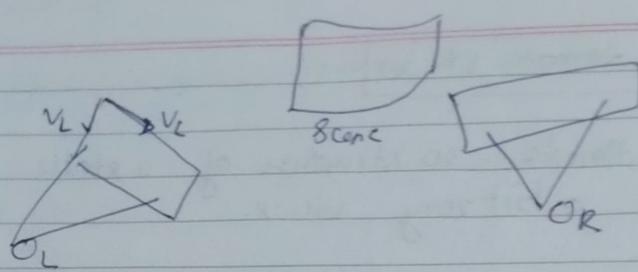
$$\begin{bmatrix} x_L \\ y_L \\ z_L \end{bmatrix} = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} \begin{bmatrix} x_r \\ y_r \\ z_r \end{bmatrix} + \begin{bmatrix} t_x \\ t_y \\ t_z \end{bmatrix} \quad \text{--- (2)}$$

Substituting eq 2 into 1. essential m.

$$\begin{bmatrix} x_L & y_L & z_L \end{bmatrix} \left(\begin{bmatrix} 0 & -t_z & ty \\ t_z & 0 & -tx \\ -ty & tx & 0 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} \begin{bmatrix} x_r \\ y_r \\ z_r \end{bmatrix} \right) = 0$$

$$\begin{bmatrix} 0 & -t_z & ty \\ t_z & 0 & -tx \\ -ty & tx & 0 \end{bmatrix} \begin{bmatrix} t_x \\ t_y \\ t_z \end{bmatrix} = 0$$

$$\boxed{t_x t = 0}$$



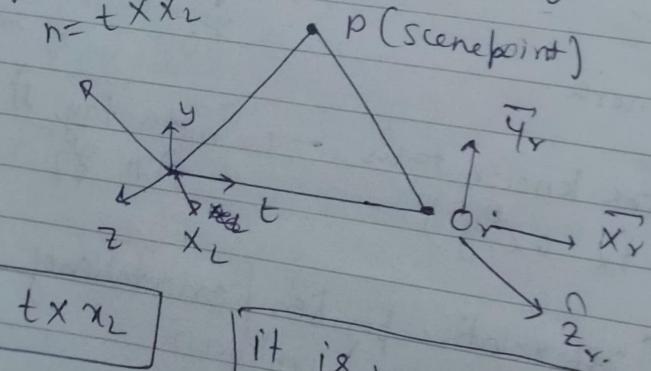
- Camera matrix (K) is known to us
- Find the small number of reliable corresponding points
(at least 8) using SIFT
↳ hand picked
- Find Relative camera position t and orientation R .

Find dense correspondence (map)
↳ compute Depth using Triangulation

Epipolar Geometry

↳ Epipole → Image point of origin / pinhole of one camera as viewed by the other camera.

$$\text{normal } n = t \times x_L$$



$$n = t \times x_L$$

it is normal to x_L

using RHT

$$\text{therefore } x_L \cdot (t \times x_L) = 0$$

↳ Epipolar constraint

Substituting

$[x_L \ y_L]$

$+ [$

we don't have x_L and x_R but we have corresponding points in image coordinates

Rewriting in terms of image coordinates

$$\begin{bmatrix} u_L & v_L & 1 \end{bmatrix} z_L K_1^{-T} \begin{bmatrix} e_{11} & e_{12} & e_{13} \\ e_{21} & e_{22} & e_{23} \\ e_{31} & e_{32} & e_{33} \end{bmatrix} K_2^{-1} z_R \begin{bmatrix} u_R \\ v_R \\ 1 \end{bmatrix} = 0$$

$\boxed{z_L \text{ can't be } 0}$ \rightarrow we can ignore this

This 3×3 matrix is also called fundamental matrix

$$\boxed{E = K_1^{-T} F K_2}$$

$$\boxed{E = T_X R}$$

$$\begin{bmatrix} x_L & y_L & z_L \end{bmatrix} \begin{bmatrix} e_{11} & e_{12} & e_{13} \\ e_{21} & e_{22} & e_{23} \\ e_{31} & e_{32} & e_{33} \end{bmatrix} \begin{bmatrix} x_r \\ y_r \\ z_r \end{bmatrix} = 0$$

↓
essential matrix

$$E = T_x R$$

Longuet Huggins — 1981 theorem

You can decompose it in T_x and R

$$\begin{bmatrix} e_{11} & e_{12} & e_{13} \\ e_{21} & e_{22} & e_{23} \\ e_{31} & e_{32} & e_{33} \end{bmatrix} = \begin{bmatrix} 0 & -t_2 & t_y \\ t_2 & 0 & -t_x \\ -t_y & t_x & 0 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix}$$

(skew)

and $R \rightarrow$ orthonormal

$$(AA^T) = I$$

it is possible to "decouple T_x and R " from their product using Single Value Decomposition

$$x_L^T E x_r = 0$$

Epipolar c.

$$\begin{bmatrix} x_L & y_L & z_L \end{bmatrix} \begin{bmatrix} e_{11} & e_{12} & e_{13} \\ e_{21} & e_{22} & e_{23} \\ e_{31} & e_{32} & e_{33} \end{bmatrix} \begin{bmatrix} x_r \\ y_r \\ z_r \end{bmatrix} = 0$$

↓
essential matrix

3D position
in right camera