

LOST 1012

Wednesday, October 12, 2022 5:15 PM

LOST SECTION 10/12

TODAY'S AGENDA:

1. Image Formation
2. Two-view reconstruction + demo

Image Formation:

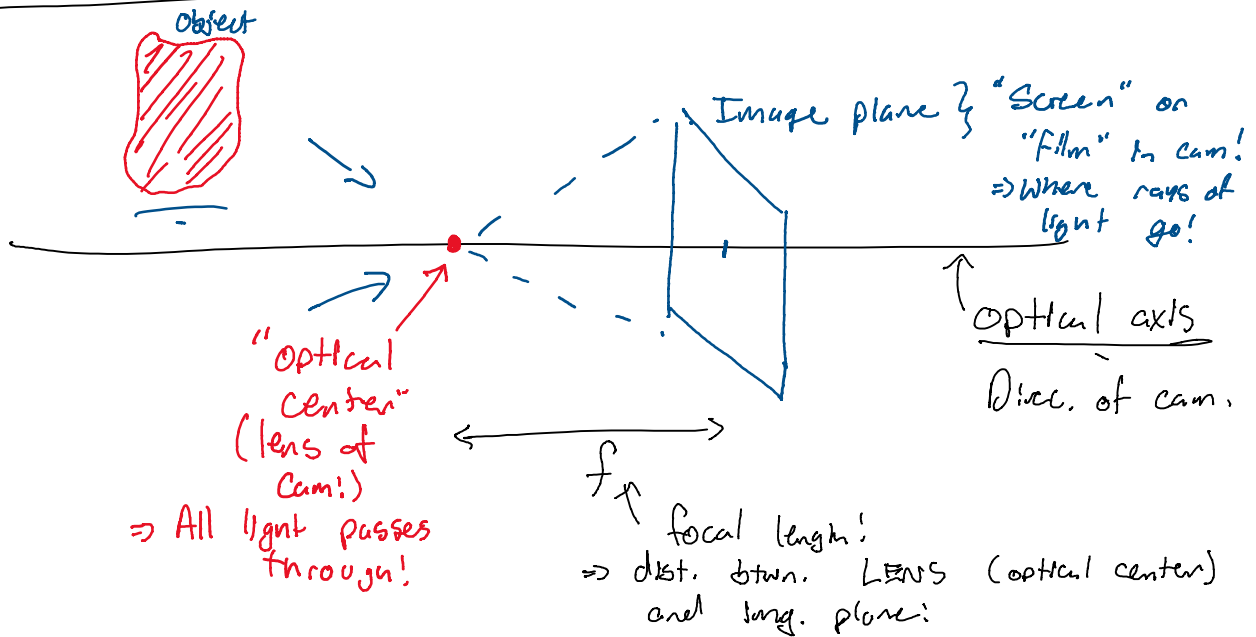
Q: how are images formed?

⇒ Come up w/ a CAM MODEL!

- Complexity: Complex models for cameras are impractical!

⇒ want to find something w/ a BALANCE of simplicity + accuracy!

PINHOLE CAMERA MODEL:

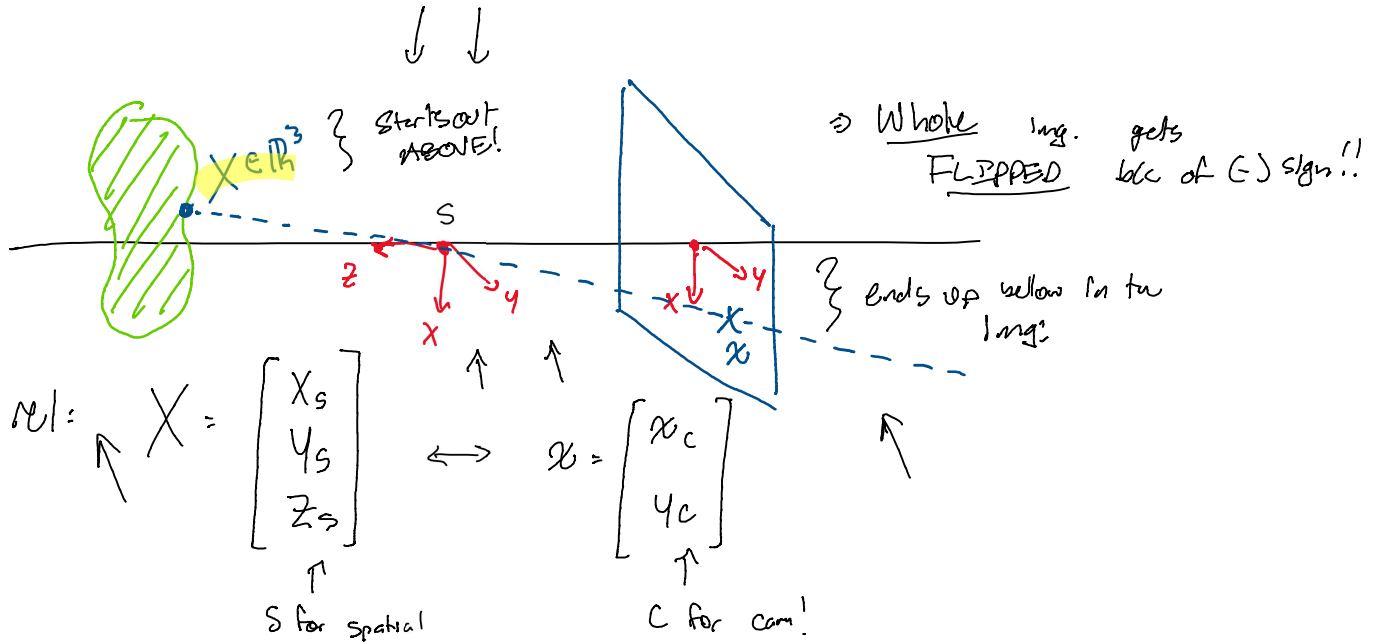
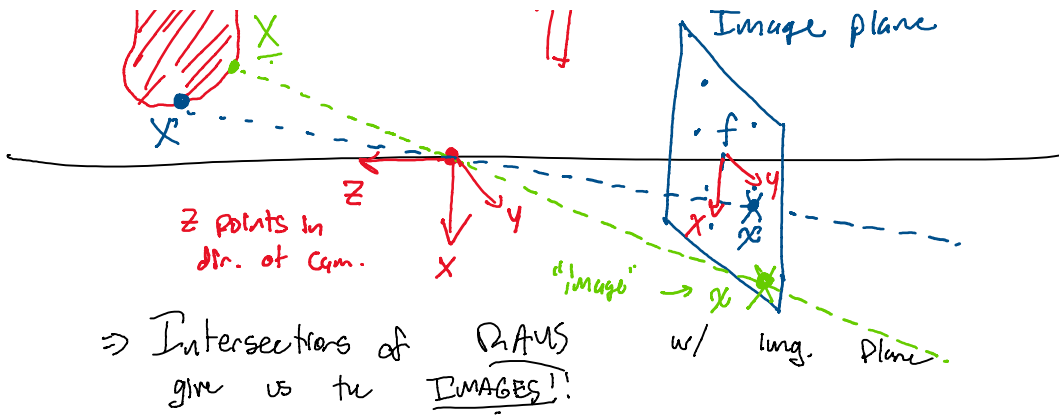


- How does light interact w/ img. plane?

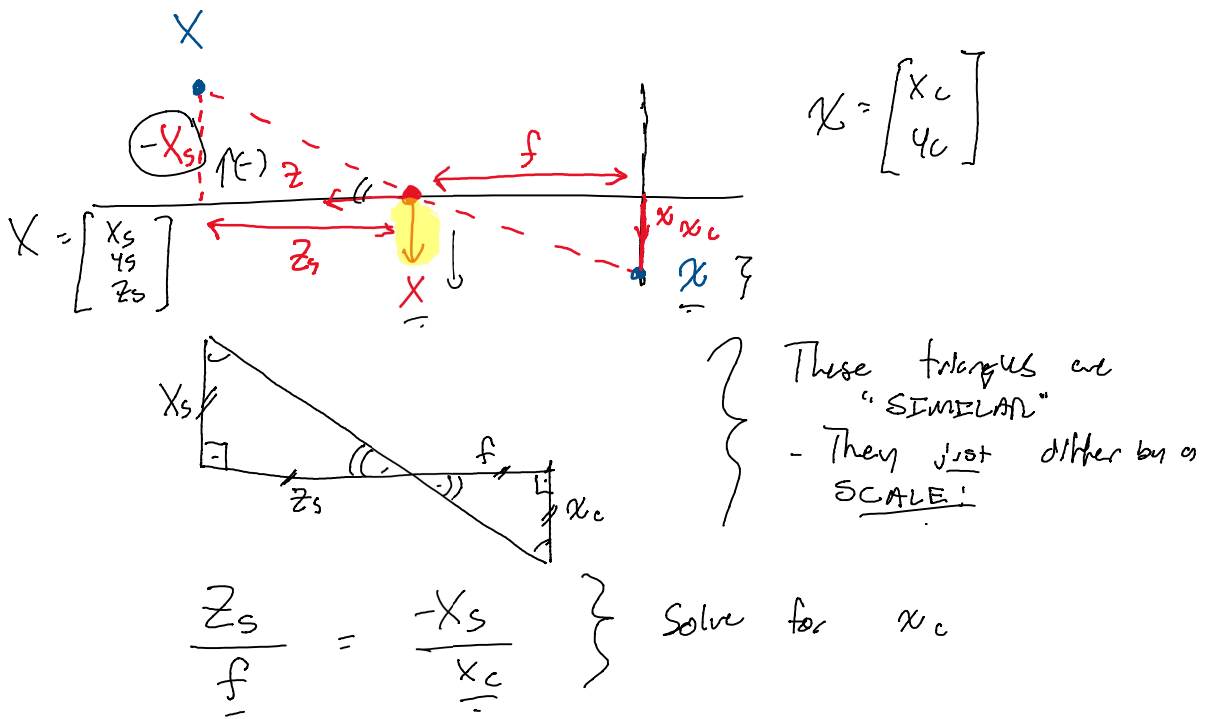


⇒

Matn. model?
⇒ Relate $X \rightarrow x$?



Try looking for a GEOM. rel!



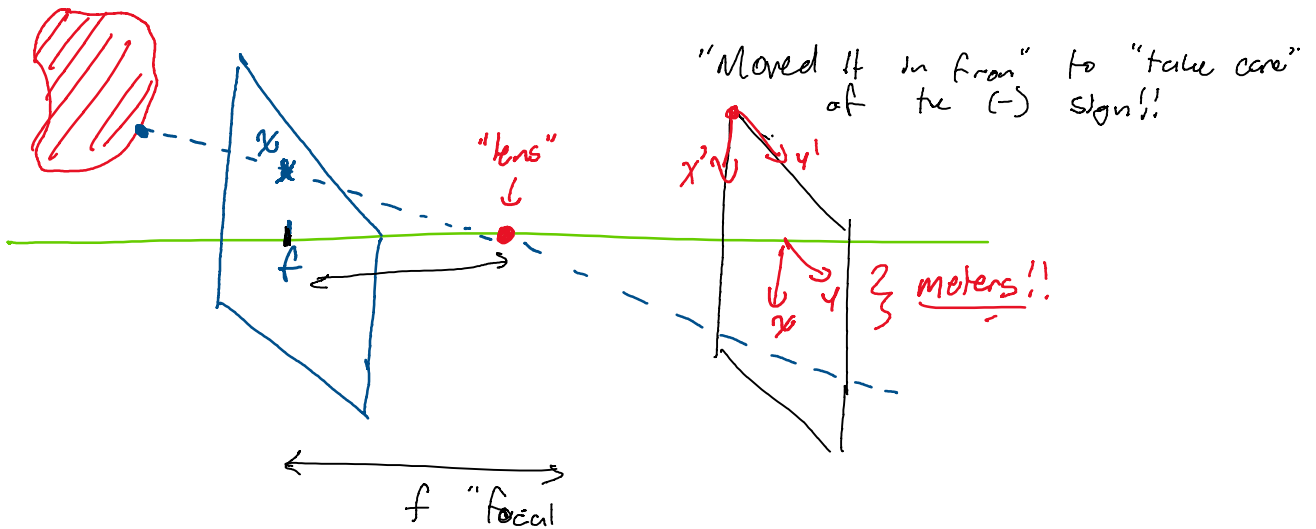
$$z_s x_c = -f x_s$$

$$\left. \begin{aligned} x_c &= \frac{-f x_s}{z_s} \\ y_c &= \frac{-f y_s}{z_s} \end{aligned} \right\} \begin{array}{l} \text{Pixel coord.} \\ \text{ENTIRELY in terms of} \\ \text{WORLD coords!} \end{array}$$

\Rightarrow Pixel locations!!

\Rightarrow If these (-) signs are a problem, let's just drop them!!

$$x_c = \frac{f x_s}{z_s} \quad y_c = \frac{f y_s}{z_s}$$



"How can we add matrices"

$$X \rightarrow x \quad \left. \begin{array}{l} \\ \end{array} \right\} \text{in matrix form!}$$

\Rightarrow Rewrite in HOMOGENEOUS coords!

homog. \rightarrow

$$X = \begin{bmatrix} x_s \\ y_s \\ z_s \\ 1 \end{bmatrix} \in \mathbb{R}^4$$

$$\bar{x} = \begin{bmatrix} x_c \\ y_c \\ 1 \end{bmatrix} \in \mathbb{R}^3$$

- BREAK our prob. into parts!

1. Reshape \bar{X} to the size of \bar{x}
2. Change 3D homog. vec. into IM6. coords!

① Reshape \bar{X} :

$$\begin{matrix} \begin{bmatrix} X_s \\ Y_s \\ Z_s \end{bmatrix} \\ 3 \times 1 \end{matrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} X_s \\ Y_s \\ Z_s \\ 1 \end{bmatrix} \begin{matrix} \left. \vphantom{\begin{bmatrix} X_s \\ Y_s \\ Z_s \\ 1 \end{bmatrix}} \right\} \text{Stay the same!} \\ \left. \vphantom{\begin{bmatrix} X_s \\ Y_s \\ Z_s \\ 1 \end{bmatrix}} \right\} \text{go to zero!} \end{matrix}$$

\uparrow
 Π_0
 "P1 = projection"

② SCALE into our \bar{x} img. coords!

$$Z_s \begin{bmatrix} x_c \\ y_c \\ 1 \end{bmatrix} = \begin{bmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} X_s \\ Y_s \\ Z_s \end{bmatrix}$$

$$x_c = \frac{f X_s}{Z_s}$$

$$y_c = \frac{f Y_s}{Z_s}$$

\uparrow
 Very clean matrix
 by mult. by Z_s
 \Rightarrow focal calibration matrix!

$$Z_s x_c = f X_s$$

$$Z_s y_c = f Y_s$$

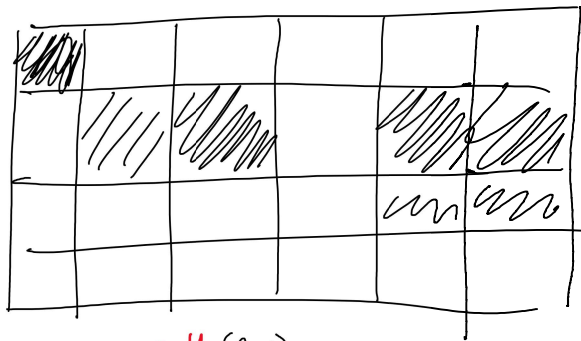
Spatial \equiv

$$Z_s \begin{bmatrix} x_c \\ y_c \\ 1 \end{bmatrix} = \begin{bmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} X_s \\ Y_s \\ Z_s \\ 1 \end{bmatrix}$$

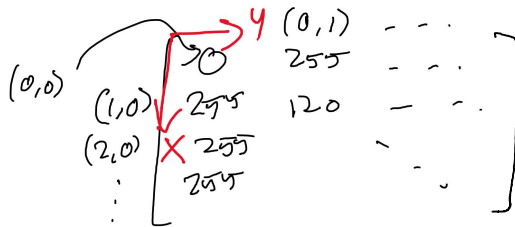
$$\boxed{Z_s \bar{x} = K_f \Pi_0 \bar{X}} \quad \star \text{ one repr. of our pinhole cam! } \star$$

$$= K_f X_s \begin{bmatrix} X_s \\ Y_s \\ Z_s \end{bmatrix}$$

- let's discuss DIGITAL images!



⇒ Image MATRIX = "intensity" values!



X & y are your row & col!
⇒ one PIXEL coords!

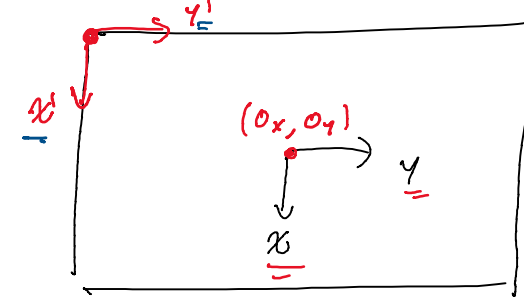
1. Change to ORIGIN

2. Scale our coords. into PIXELS (from m, mm, ...)

↑
"Scaling Calibration matrix"

$$K_s = \begin{bmatrix} S_x & S_y & 0_x \\ 0 & 0 & 0_y \\ 0 & 0 & 0 \end{bmatrix}$$

→ MOVE the origin of our image!



Complete transf. into PIXEL coords:

$$Z_s \bar{x} = K_s K_f \Pi_0 \bar{X}$$

↑ ↑ ↑
Scale focal reshape (project)

$K = K_s K_f$ "CAMERA CALIBRATION MATRIX"

Spatial points! 480 x 720 img
 $0_x = 240$
 $0_y = 360$

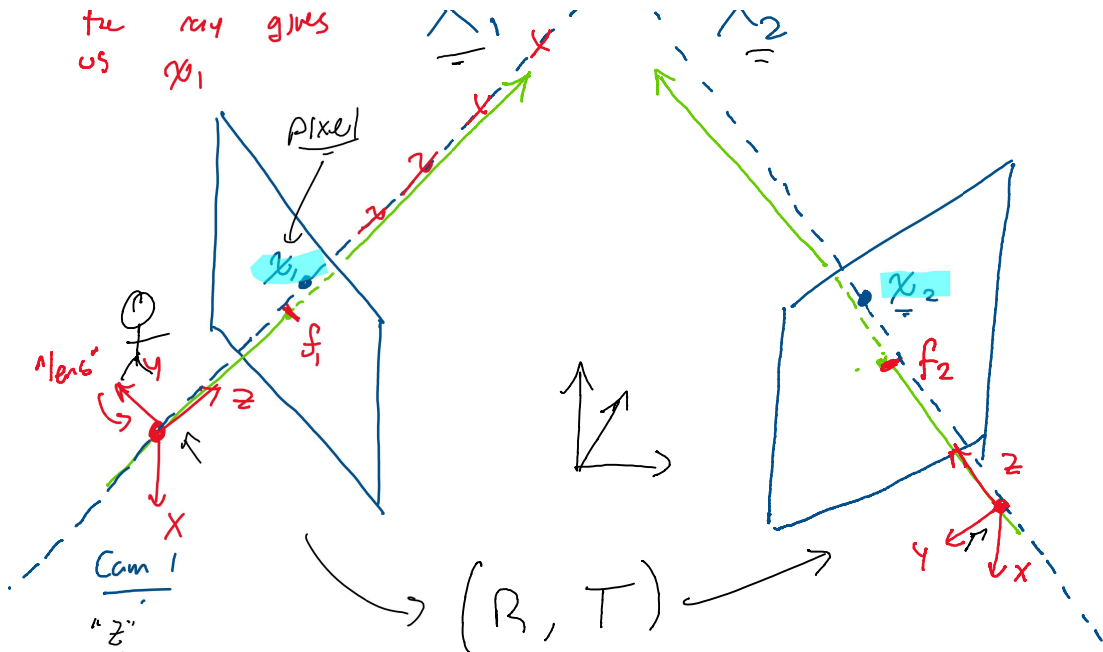
TWO VIEW RECONSTR:

⇒ RECOVER DEPTH FROM IMAGES!
⇒ Important for robotics! ex: obstacles!

⇒ Epipolar constr + Essential matrix!

Part I: Recovering Depth:

⇒ Everything along



What we know!

x_1, x_2

What we want

\rightarrow either X_1 or X_2

$$\lambda_1 \bar{x}_1 = K_1 \Pi_0 \bar{X}_1$$

$$\boxed{\lambda_1 \bar{x}_1 = K_1 X_1}$$

$$\boxed{\lambda_2 \bar{x}_2 = K_2 X_2}$$

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = \begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$$

sets "1" in homography coords to 0!

$$\begin{bmatrix} X_1 \\ Y_1 \\ Z_1 \end{bmatrix}$$

$$\lambda_1 \bar{x}_1 = K_1 X_1$$

$$\lambda_1 K_1^{-1} \bar{x}_1 = X_1$$

"Normalized coords"

$$\underline{x_1 = K_1^{-1} \bar{x}_1}$$

$$\left\{ \begin{array}{l} \lambda_1 \bar{x}_1 = X_1 \\ \lambda_2 \bar{x}_2 = X_2 \end{array} \right\}$$

$$X_1 = \begin{bmatrix} X_s \\ Y_s \\ Z_s \end{bmatrix}$$

How can we RELATE x_1 & x_2 ?

$$X_2 = R X_1 + T$$

$$\lambda_2 \bar{x}_2 = R (\lambda_1 \bar{x}_1) + T$$

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

} ONLY things we DON'T know λ_1, λ_2

$$-\lambda_1 R \bar{x}_1 + \lambda_2 \bar{x}_2 = T$$

$$v = \begin{bmatrix} \lambda_1 \\ \lambda_2 \end{bmatrix}$$

$$A v = b \quad \underline{v = A^{-1} b}$$

\uparrow \uparrow \uparrow
 Matrix! Vector of known!
 (unknown) unknowns

$\begin{bmatrix} \lambda_1 \\ \lambda_2 \end{bmatrix}$

$$\begin{bmatrix} | & | \\ -R x_1 & x_2 \\ | & | \end{bmatrix} \begin{bmatrix} \lambda_1 \\ \lambda_2 \end{bmatrix} = T$$

$\left. \begin{array}{l} \text{"A"} \\ 3 \times 1 \text{ vec!} \quad 3 \times 1 \text{ vec} \end{array} \right\} \text{Can solve for } \lambda_1 \text{ \& } \lambda_2!$

$A \in \mathbb{R}^{3 \times 2}$ matrix!

"Pseudo inverse" \Rightarrow "Moore-Penrose pseudo inv"

$A^+ = (A^T A)^{-1} A^T$ } Pseudo inv. of A

$A^+ A = (A^T A)^{-1} A^T \cdot A = (A^T A)^{-1} (A^T A) = I$

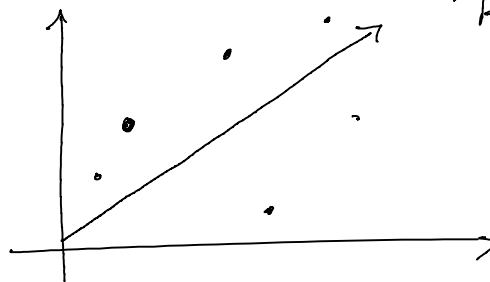
~~$$\begin{bmatrix} -R x_1 & x_2 \\ | & | \end{bmatrix}^+ \begin{bmatrix} -R x_1 & x_2 \\ | & | \end{bmatrix} \begin{bmatrix} \lambda_1 \\ \lambda_2 \end{bmatrix} = \begin{bmatrix} -R x_1 & x_2 \\ | & | \end{bmatrix}^T$$~~

$$\begin{bmatrix} \lambda_1 \\ \lambda_2 \end{bmatrix} = \begin{bmatrix} -R x_1 & x_2 \\ | & | \end{bmatrix}^+ T$$

\Rightarrow gives us our depths!!

$$\left. \begin{array}{l} \lambda_1 x_1 = X_1 \\ \lambda_2 x_2 = X_2 \end{array} \right\} \text{Both part to same physical pt.}$$

"Least sq. soln"



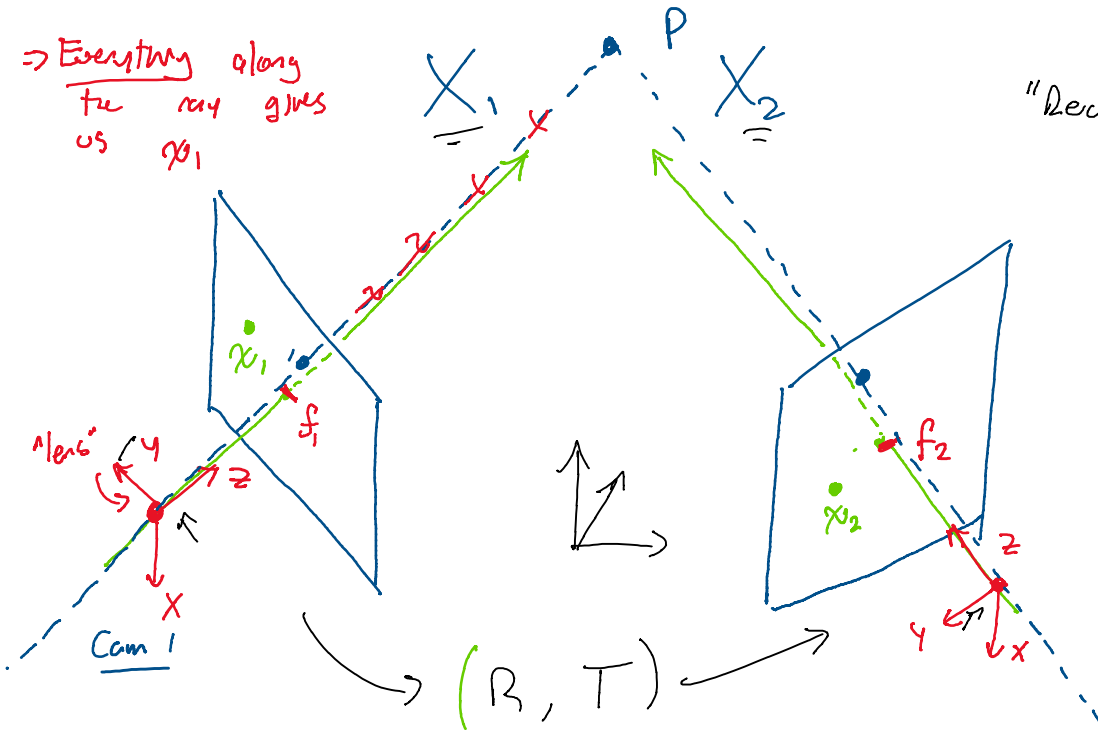
"pseudo inv" gives BEST soln. for that case.

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ESSENTIAL MATRIX:

→ Everything along the ray gives us x_1

"Reconstruction"



How do we know if x_1, x_2 (pixels) POINT to the same 3D SPATIAL location??

(from two images!)

$$\underline{x}_2 = R \underline{x}_1 + T$$

$$\lambda_2 x_2 = \lambda_1 R x_1 + \underline{T}$$

⇒ REDUCE to the SIMPLEST constr. possible!

- elim. unknowns! (λ_1, λ_2)
- eliminate unrec. terms.

$$\underline{T}^T x \cdot v = \hat{T}^T v$$

matrix!!

$$\underline{T}^T \times \underline{T} = 0$$

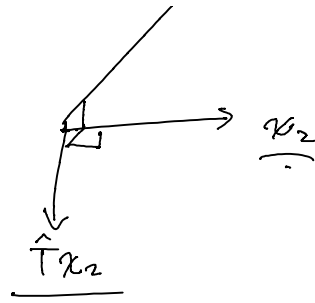
$$\lambda_2 \hat{T}^T x_2 = \lambda_1 \hat{T}^T R x_1 + \cancel{\hat{T}^T T} = 0$$

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

$$\hat{T}^T$$

$$x \perp y$$

$$x \cdot y = x^T y$$



$$x \cdot y = 0$$

$$\begin{aligned} x \cdot y &= \|x\| \|y\| \cos \theta \\ &= \|x\| \|y\| \cos(90^\circ) \\ &= 0 \end{aligned}$$

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

$$0 = \lambda_1 x_2^T \hat{T} R x_1$$

$$\boxed{0 = x_2^T \hat{T} R x_1} \quad \text{"EPIPOLAR CONSTR!"}$$

⇒ If x_2 & x_1 corr. to the SAME pt. in space:

$$x_2^T \hat{T} R x_1 = 0$$

⇒ "ESSENTIAL MATRIX" $\boxed{E = \hat{T} R}$

$$x_2^T E x_1 = 0$$

If x_2 & x_1 are from the same physical location.

x_1, x_2, \dots, x_n

$$\underline{X} E^S = 0$$

X : Matrix that encodes information about x_1, \dots, x_n

$q \times q$ "Kronecker product" \otimes

$$E^S = ?? \quad \underline{E^S} = \begin{bmatrix} e_{11} \\ e_{21} \\ e_{31} \\ \vdots \\ e_{33} \end{bmatrix} \in \mathbb{R}^q \quad E = \begin{bmatrix} e_{11} & \dots \\ e_{21} & \dots \\ \vdots & \vdots \\ e_{33} \end{bmatrix}$$

↑
don't know, want to recover!!

$$X E^S = 0$$

$\Rightarrow E^s$ should be UNIQUE up to a scale factor!!

\Rightarrow If E^s ONLY points in a single dir, \mathcal{X} is a 9×9 matrix:

\Rightarrow Null space of \mathcal{X} to be (10)

\Rightarrow Column space of \mathcal{X} is (80)

- MUST PICK (8) pts. to get \mathcal{X} a 10 null space, when gives $E^s \Rightarrow E$ up to a scale factor.

$$\lambda \cdot T_{..} + \lambda \cdot T_{..} + \lambda \cdot T_{..} = 0$$

$$\begin{bmatrix} \lambda & \lambda & \lambda \\ \dots & \dots & \dots \\ \lambda & \lambda & \lambda \end{bmatrix} \begin{bmatrix} \lambda_1 \\ \lambda_2 \\ \lambda_3 \end{bmatrix} = 0$$

$$\text{Pt cloud} = \begin{bmatrix} \lambda_1 & \dots & \lambda_n \\ \lambda_1 & \dots & \lambda_n \\ \lambda_1 & \dots & \lambda_n \end{bmatrix}$$

$$\| \text{pt cloud} \| = [\|x_1\| \dots \|x_n\|]$$

np. argmin ($\lambda \dots \|$)

$$U_{sn} = \begin{bmatrix} \lambda_j & \lambda_{s+1} \\ \lambda_j & \lambda_{s+1} \end{bmatrix}$$