



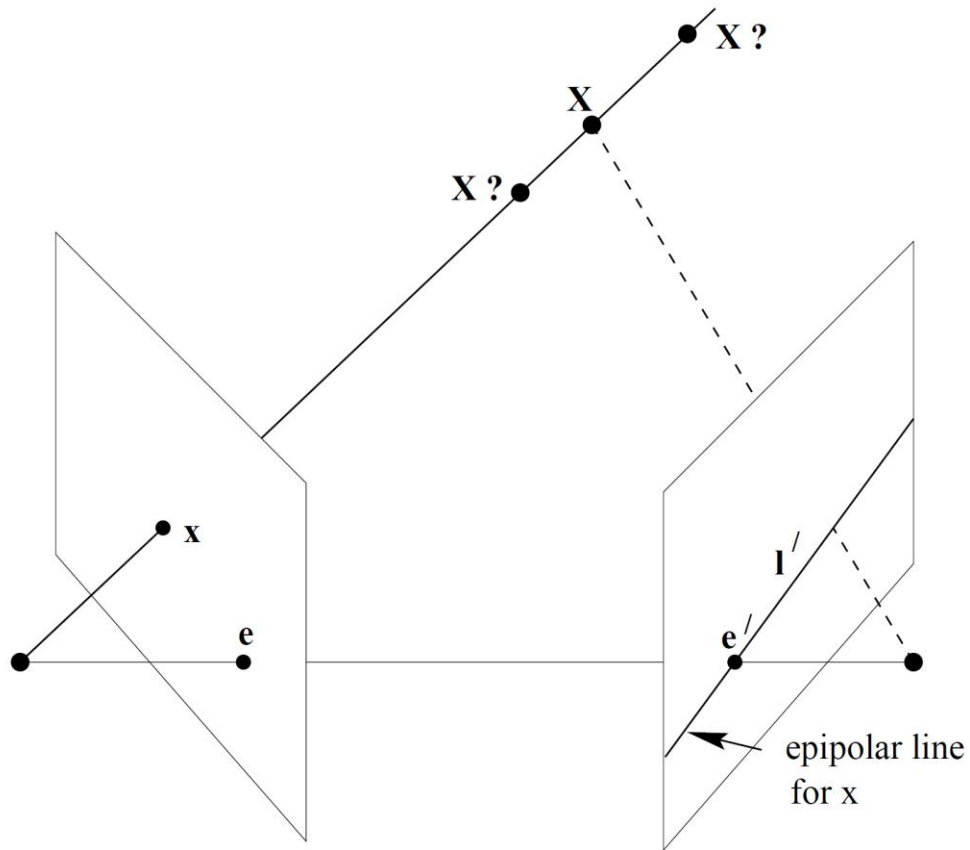
# Epipolar Geometry and Stereo

CS 6384 Computer Vision

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# Recall Fundamental Matrix



- Epipolar line  $\mathbf{l}' = F \mathbf{x}$   
 $\mathbf{l} = F^T \mathbf{x}'$

- Fundamental matrix

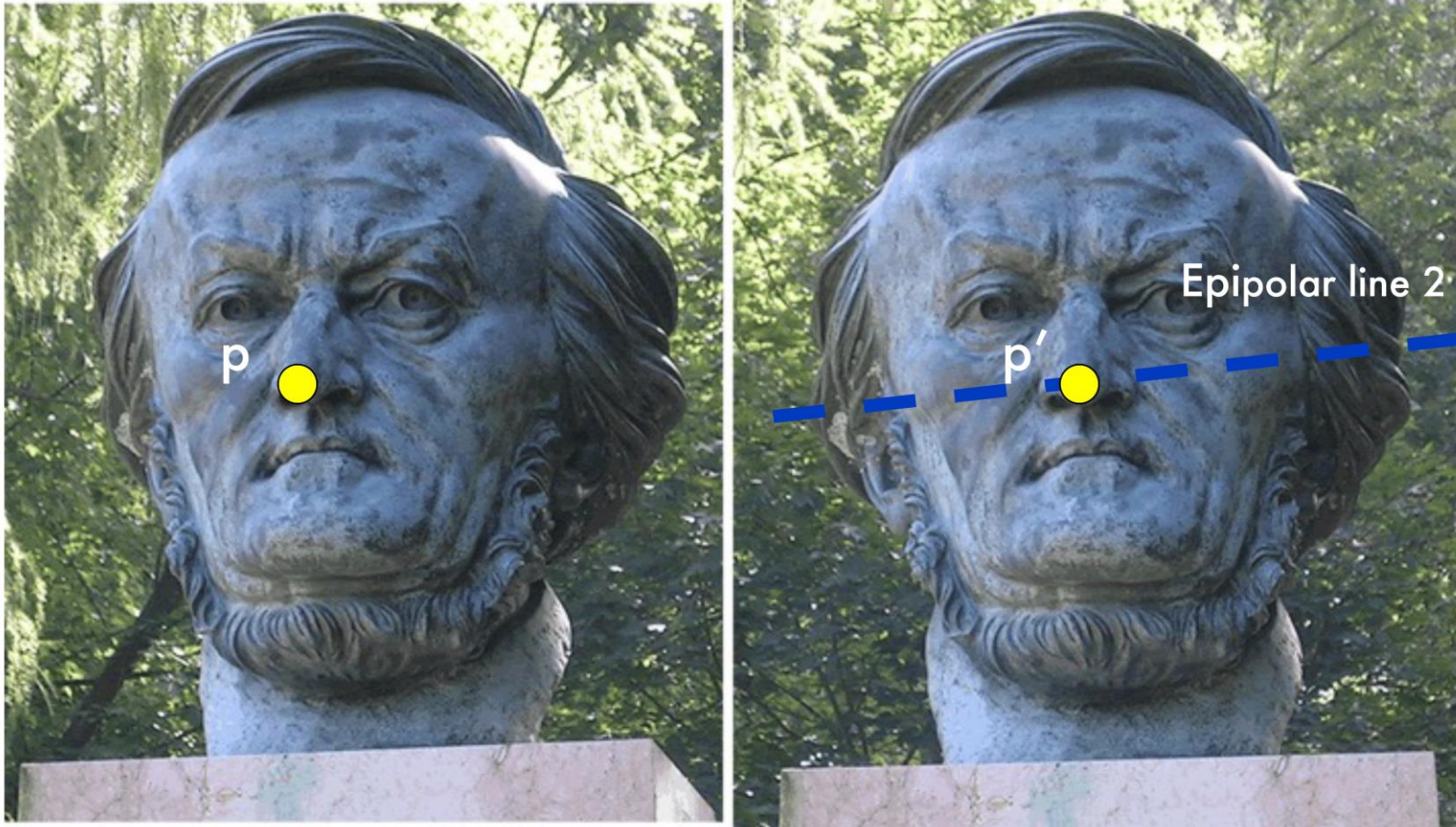
$$F = [\mathbf{e}']_{\times} P' P^+$$

3x3

Epipole  $\mathbf{e}' = (P' C)$

$$P^+ = P^T (P P^T)^{-1}$$

# Why the Fundamental Matrix is Useful?

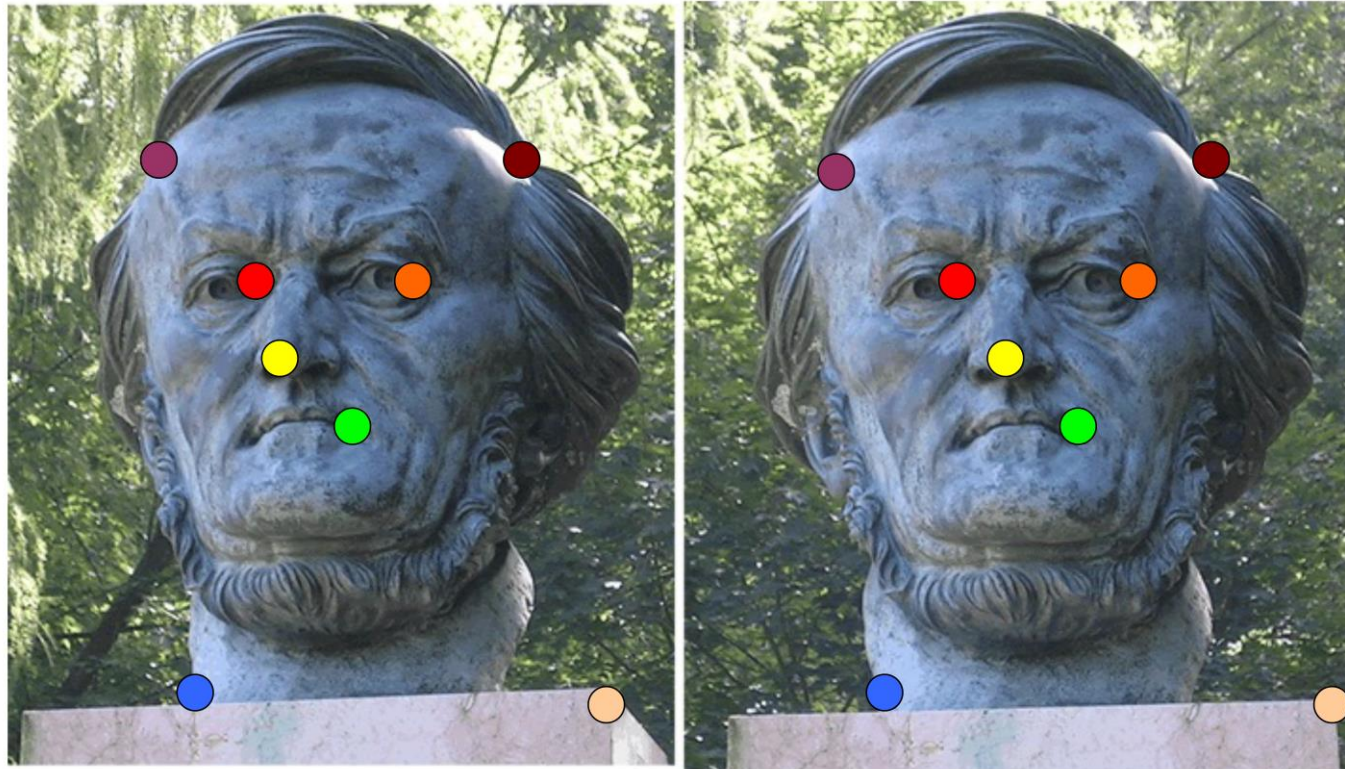


$$l' = Fp$$



# Estimating the Fundamental Matrix

- The 8-point algorithm



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

# Estimating the Fundamental Matrix

$$\mathbf{x}'^\top \mathbf{F} \mathbf{x} = 0 \quad \mathbf{x} = (x, y, 1)^\top \quad \mathbf{x}' = (x', y', 1)^\top$$

$$x'x f_{11} + x'y f_{12} + x' f_{13} + y'x f_{21} + y'y f_{22} + y' f_{23} + x f_{31} + y f_{32} + f_{33} = 0$$

$$(x'x, x'y, x', y'x, y'y, y', x, y, 1) \mathbf{f} = 0$$

n correspondences

$$\mathbf{A} \mathbf{f} = \begin{bmatrix} x'_1 x_1 & x'_1 y_1 & x'_1 & y'_1 x_1 & y'_1 y_1 & y'_1 & x_1 & y_1 & 1 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ x'_n x_n & x'_n y_n & x'_n & y'_n x_n & y'_n y_n & y'_n & x_n & y_n & 1 \end{bmatrix} \mathbf{f} = \mathbf{0}$$

# Linear System

$$\begin{matrix} & A\mathbf{f} = 0 \\ n \times 9 & 9 \times 1 \end{matrix}$$

- Find non-zero solutions
- If  $\mathbf{f}$  is a solution,  $k\mathbf{f}$  is also a solution for  $k \in \mathcal{R}$
- If the rank of  $A$  is 8, unique solution (up to scale)
- Otherwise, we can seek a solution  $\|\mathbf{f}\| = 1$

$\min \|\mathbf{A}\mathbf{f}\|$   
Subject to  $\|\mathbf{f}\| = 1$

Solution:  $A = UDV^T$  SVD decomposition of  $A$

$n \times n \quad n \times 9 \quad 9 \times 9$

$\mathbf{f}$  is the last column of  $V$

A5.3 in HZ

# Estimating the Fundamental Matrix

- The singularity constraint  $\det F = 0$

$$\begin{aligned} & \min \|F - F'\| \\ & \text{Subject to } \det F' = 0 \end{aligned}$$

$$F = UDV^T$$

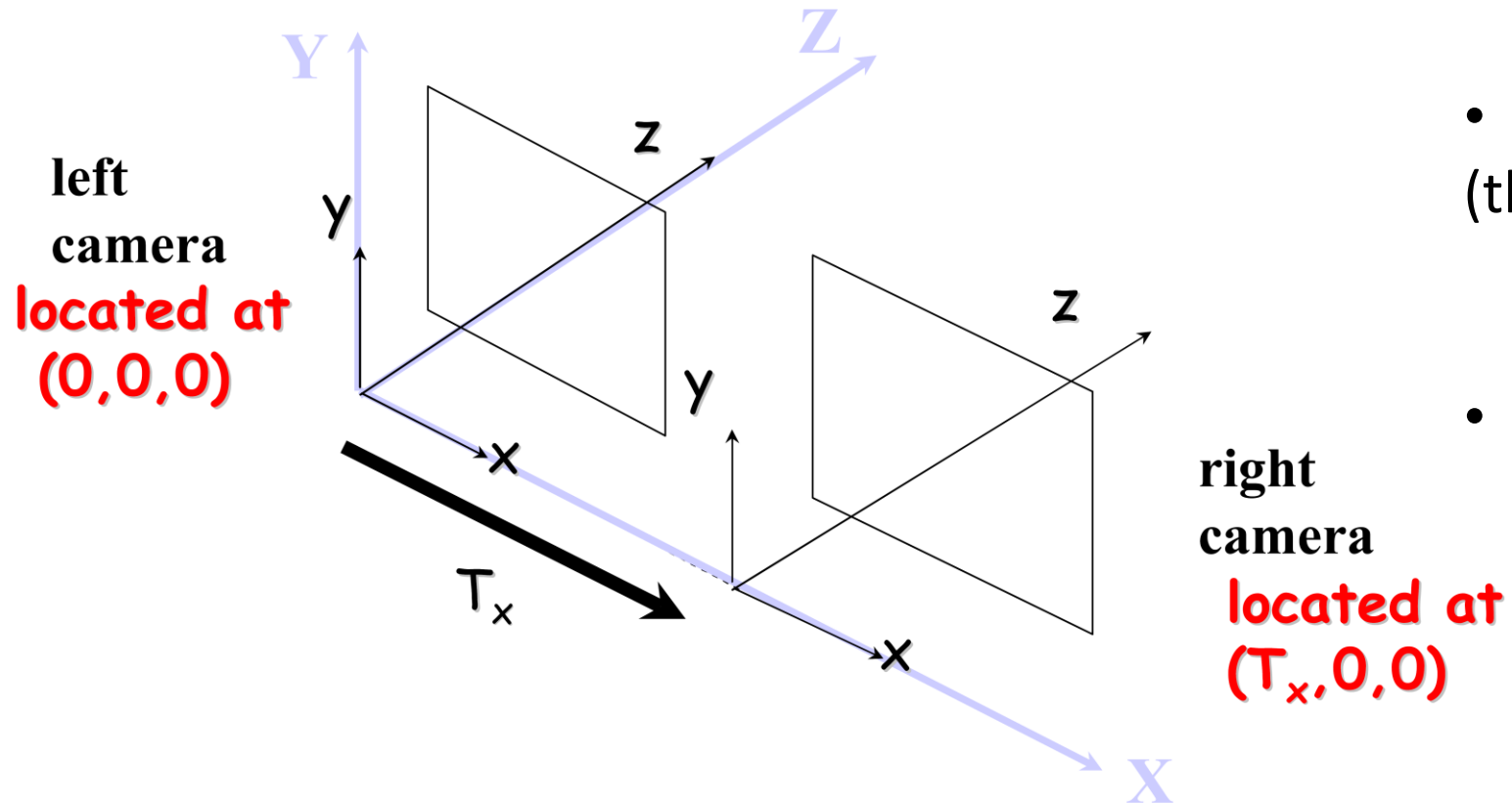
Solution:

$$D = \text{diag}(r, s, t)$$

$$F' = U \text{diag}(r, s, 0) V^T$$

$$r \geq s \geq t$$

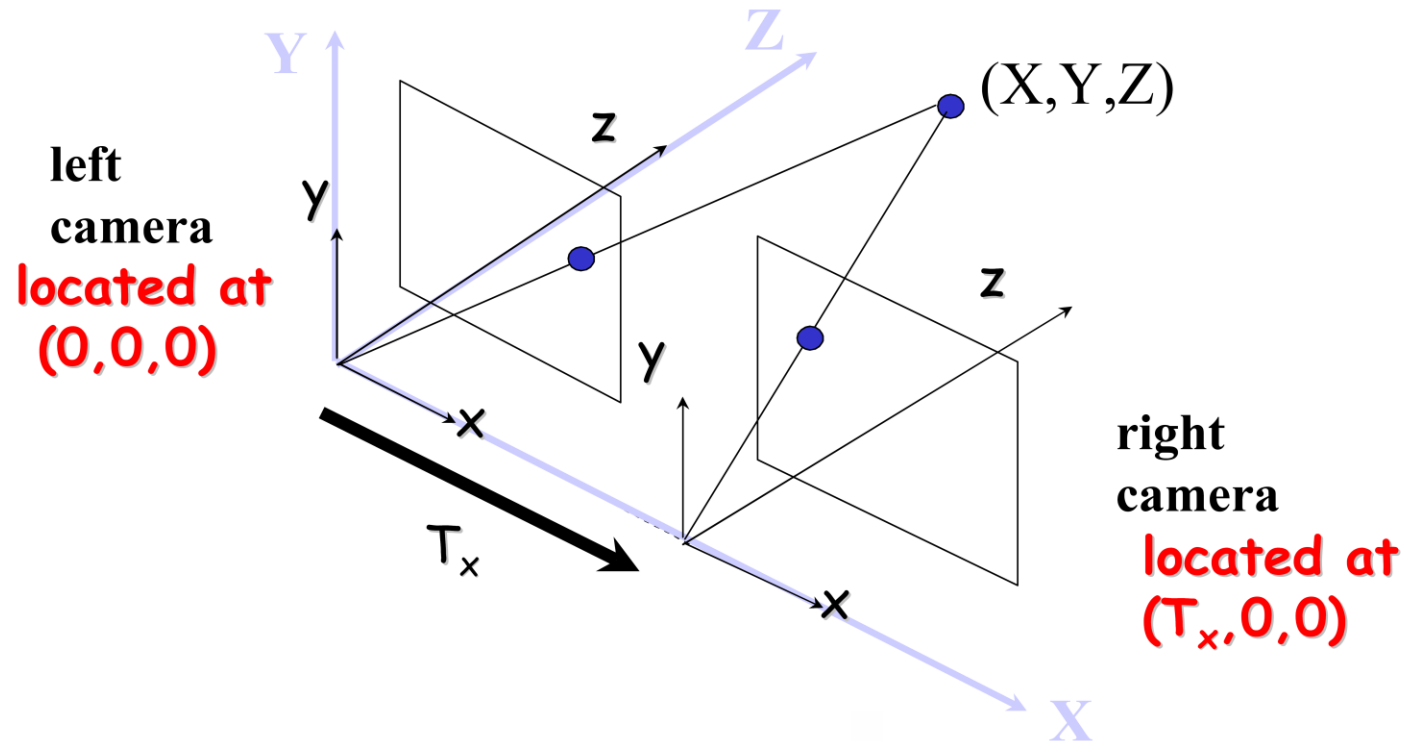
# Special Case: A Stereo System



- The right camera is shifted by  $T_x$  (the stereo baseline)
- The camera intrinsics are the same



# Special Case: A Stereo System



- Left camera

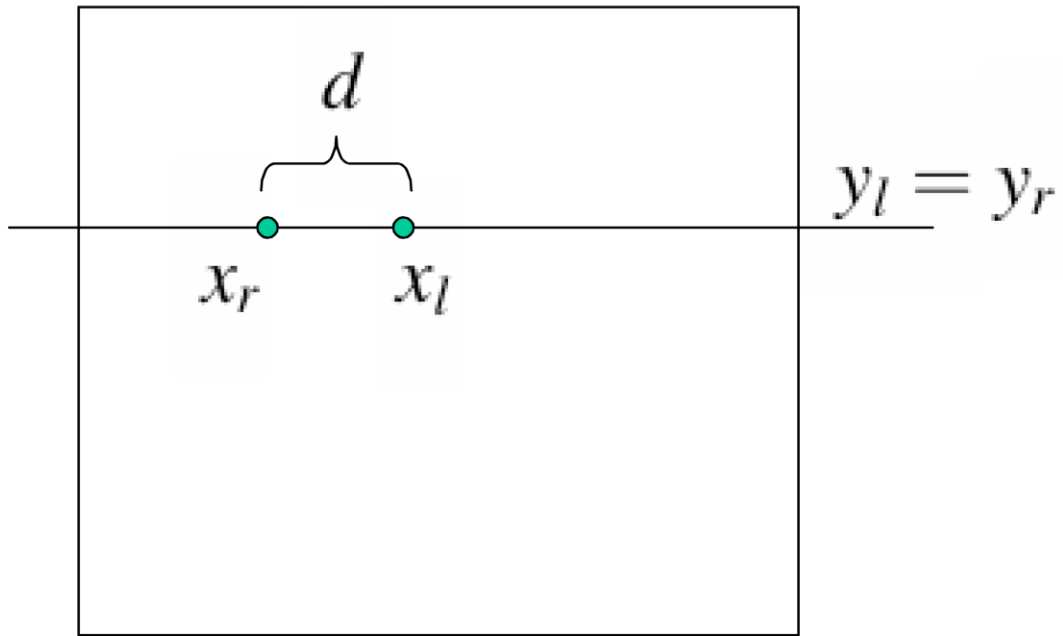
$$x_l = f \frac{X}{Z} + p_x \quad y_l = f \frac{Y}{Z} + p_y$$

- Right camera

$$x_r = f \frac{X - T_x}{Z} + p_x$$

$$y_r = f \frac{Y}{Z} + p_y$$

# Stereo Disparity



- Disparity

$$\begin{aligned}d &= x_l - x_r \\&= \left(f \frac{X}{Z} + p_x\right) - \left(f \frac{X - T_x}{Z} + p_x\right) \\&= f \frac{T_x}{Z}\end{aligned}$$

- Depth

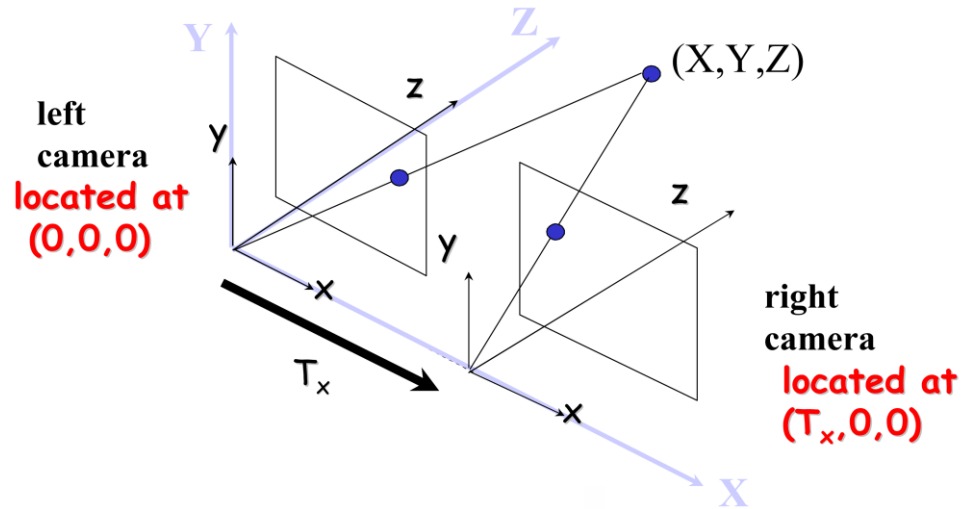
$$Z = f \frac{T_x}{d}$$

Baseline

Disparity

Recall motion parallax: near objects move faster (large disparity)

# Special Case: A Stereo System



$$P = K[I \mid \mathbf{0}] \quad P' = K[I \mid \mathbf{t}]$$

$$F = [\mathbf{e}']_{\times} K' R K^{-1} = K'^{-T} [\mathbf{t}]_{\times} R K^{-1} = K'^{-T} R [R^T \mathbf{t}]_{\times} K^{-1} = K'^{-T} R K^T [\mathbf{e}]_{\times}$$

$$F = [\mathbf{e}']_{\times} K K^{-1} = [\mathbf{e}']_{\times}$$

$$\mathbf{e}' = (P' C) \quad C = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

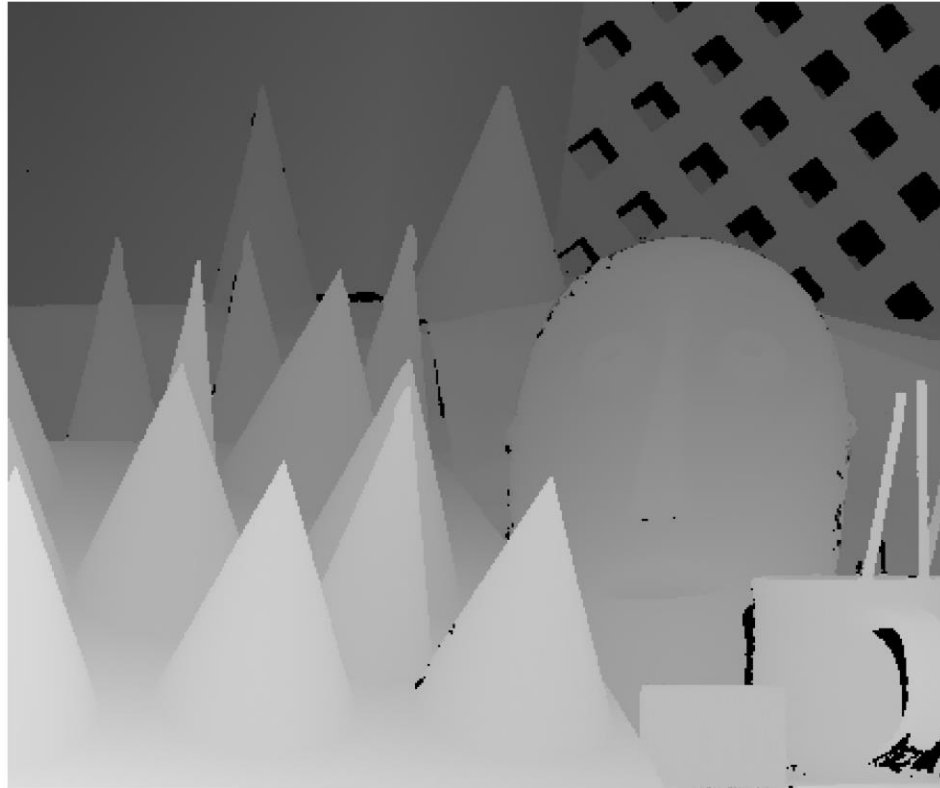
$$K = \begin{bmatrix} f_x & 0 & p_x \\ 0 & f_y & p_y \\ 0 & 0 & 1 \end{bmatrix} \quad \mathbf{e}' = \begin{bmatrix} f_x T_x \\ 0 \\ 0 \end{bmatrix}$$

$$F = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & -f_x T_x \\ 0 & f_x T_x & 0 \end{bmatrix} \quad \begin{aligned} \mathbf{x}'^T F \mathbf{x} &= 0 \\ y &= y' \end{aligned}$$

# Stereo Example



Disparity values (0-64)



Note how disparity is larger (brighter) for closer surfaces.

$$d = f \frac{T_x}{Z}$$

# Computing Disparity

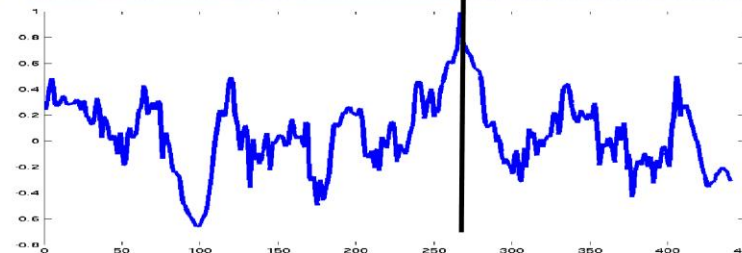
Left Image



Right Image



For a patch in left image  
Compare with patches along  
same row in right image



Match Score Values

- Eipipolar lines are horizontal lines in stereo
- For general cases, we can find correspondences on eipipolar lines
- Depth from disparity

$$Z = f \frac{T_x}{d}$$

# Further Reading

- Stanford CS231A: Computer Vision, From 3D Reconstruction to Recognition, Lecture 5 <https://web.stanford.edu/class/cs231a/syllabus.html>
- Multiview Geometry in Computer Vision, Richard Hartley and Andrew Zisserman, Chapter 9, Epipolar Geometry and Fundamental Matrix