

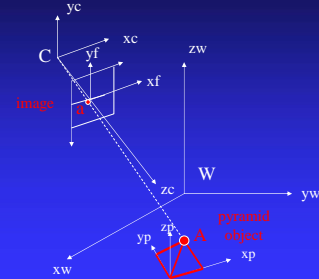
3D Sensing

- Camera Model and 3D Transformations
- Camera Calibration (Tsai's Method)
- Depth from General Stereo (overview)
- Pose Estimation from 2D Images (skip)
- 3D Reconstruction

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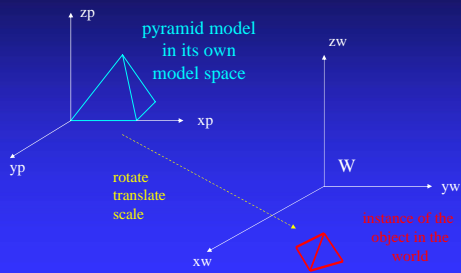
Camera Model: Recall there are 5 Different Frames of Reference

- Object
- World
- Camera
- Real Image
- Pixel Image



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Rigid Body Transformations in 3D



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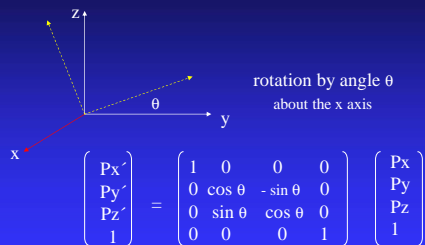
Translation and Scaling in 3D

$$\begin{bmatrix} {}^2P_x \\ {}^2P_y \\ {}^2P_z \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & x_0 \\ 0 & 1 & 0 & y_0 \\ 0 & 0 & 1 & z_0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} {}^1P_x \\ {}^1P_y \\ {}^1P_z \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} {}^2P_x \\ {}^2P_y \\ {}^2P_z \\ 1 \end{bmatrix} = \begin{bmatrix} s_x & {}^2P_x \\ s_y & {}^2P_y \\ s_z & {}^2P_z \\ 1 \end{bmatrix} = \begin{bmatrix} s_x & 0 & 0 & 0 \\ 0 & s_y & 0 & 0 \\ 0 & 0 & s_z & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} {}^1P_x \\ {}^1P_y \\ {}^1P_z \\ 1 \end{bmatrix}$$

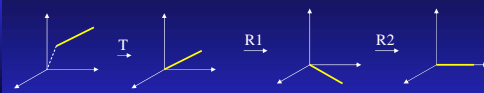
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Rotation in 3D is about an axis



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Rotation about Arbitrary Axis



One translation and two rotations to line it up with a major axis. Now rotate it about that axis. Then apply the reverse transformations (R2, R1, T) to move it back.

$$\begin{bmatrix} P_{x'} \\ P_{y'} \\ P_{z'} \\ 1 \end{bmatrix} = \begin{bmatrix} c11 & c12 & c13 & t1 \\ c21 & c22 & c23 & t2 \\ c31 & c32 & c33 & t3 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} P_x \\ P_y \\ P_z \\ 1 \end{bmatrix}$$

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The Camera Model

How do we get an **image point IP** from a **world point P**?

$$\begin{pmatrix} s \text{ Ipr} \\ s \text{ Ipc} \\ s \end{pmatrix} = \begin{pmatrix} c11 & c12 & c13 & c14 \\ c21 & c22 & c23 & c24 \\ c31 & c32 & c33 & 1 \end{pmatrix} \begin{pmatrix} P_x \\ P_y \\ P_z \\ 1 \end{pmatrix}$$

image point

camera matrix **C**

world point

What's in **C**?

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The camera model handles the **rigid body transformation** from world coordinates to camera coordinates plus the **perspective transformation** to image coordinates.

1. $CP = TR WP$
2. $IP = \pi(f) CP$

$$\begin{pmatrix} s \text{ Ipx} \\ s \text{ Ipy} \\ s \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1/f & 1 \end{pmatrix} \begin{pmatrix} CP_x \\ CP_y \\ CP_z \\ 1 \end{pmatrix}$$

image point

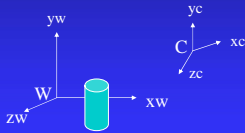
perspective transformation

3D point in camera coordinates

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Camera Calibration

- In order work in 3D, we need to know the parameters of the particular camera setup.
- Solving for the camera parameters is called calibration.

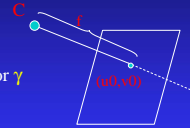


- **intrinsic** parameters are of the camera device
- **extrinsic** parameters are where the camera sits in the world

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Intrinsic Parameters

- principal point (u_0, v_0)
- scale factors (dx, dy)
- aspect ratio distortion factor γ
- focal length f
- lens distortion factor κ (models radial lens distortion)



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Extrinsic Parameters

- translation parameters $t = [t_x \ t_y \ t_z]$
- rotation matrix

$$R = \begin{pmatrix} r11 & r12 & r13 & 0 \\ r21 & r22 & r23 & 0 \\ r31 & r32 & r33 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \quad \text{Are there really nine parameters?}$$

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Calibration Object

The idea is to snap images at different depths and get a lot of **3D-2D point correspondences**.



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The Tsai Procedure

- The Tsai procedure was developed by Roger Tsai at IBM Research and is most widely used.
- Several images are taken of the calibration object yielding point correspondences at different distances.
- Tsai's algorithm requires $n > 5$ correspondences

$$\{(x_i, y_i, z_i), (u_i, v_i) \mid i = 1, \dots, n\}$$

between (real) image points and 3D points.

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In this* version of Tsai's algorithm,

- The real-valued (u, v) are computed from their pixel positions (r, c) :

$$u = \gamma dx (c - u_0) \quad v = -dy (r - v_0)$$

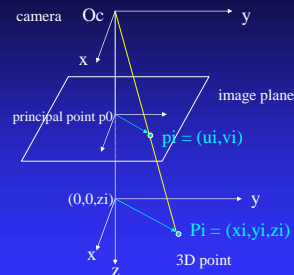
where

- (u_0, v_0) is the **center of the image**
- dx and dy are the **center-to-center (real) distances** between pixels and come from the camera's specs
- γ is a **scale factor** learned from previous trials

* This version is for single-plane calibration.

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Tsai's Geometric Setup



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Tsai's Procedure

1. Given the n point correspondences $((x_i, y_i, z_i), (u_i, v_i))$

Compute matrix A with rows a_i

$$a_i = (v_i * x_i, v_i * y_i, -u_i * x_i, -u_i * v_i, v_i)$$

These are known quantities which will be used to solve for intermediate values, which will then be used to solve for the parameters sought.

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Intermediate Unknowns

2. The vector of **unknowns** is $\mu = (\mu_1, \mu_2, \mu_3, \mu_4, \mu_5)$:

$$\mu_1 = r_{11}/t_y \quad \mu_2 = r_{12}/t_y \quad \mu_3 = r_{21}/t_y \quad \mu_4 = r_{22}/t_y \quad \mu_5 = t_x/t_y$$

where the r 's and t 's are unknown rotation and translation parameters.

3. Let vector $\mathbf{b} = (u_1, u_2, \dots, u_n)$ contain the **u image coordinates**.

4. **Solve** the system of linear equations

$$A \mu = \mathbf{b}$$

for unknown parameter vector μ .

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Use μ to solve for t_y , t_x , and 4 rotation parameters

5. Let $U = \mu_1^2 + \mu_2^2 + \mu_3^2 + \mu_4^2$. Use U to calculate t_y^2 . (see text)
6. Try the positive square root $t_y = (t_y^2)^{1/2}$ and use it to compute translation and rotation parameters.

$$\begin{aligned} r_{11} &= \mu_1 t_y \\ r_{12} &= \mu_2 t_y \\ r_{21} &= \mu_3 t_y \\ r_{22} &= \mu_4 t_y \\ t_x &= \mu_5 t_y \end{aligned}$$

Now we know
2 translation parameters and
4 rotation parameters.

except...

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Determine true sign of t_y and compute remaining rotation parameters.

7. Select an object point P whose image coordinates (u,v) are far from the image center.
8. Use P 's coordinates and the translation and rotation parameters so far to estimate the image point that corresponds to P .

If its coordinates have the same signs as (u,v) , then keep t_y , else negate it.
9. Use the first 4 rotation parameters to calculate the remaining 5.

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Solve another linear system.

10. We have t_x and t_y and the 9 rotation parameters. Next step is to find t_z and f .

Form a matrix A' whose rows are:

$$a_i' = (r_{21} * x_i + r_{22} * y_i + t_y, \quad v_i)$$

and a vector b' whose rows are:

$$b_i' = (r_{31} * x_i + r_{32} * y_i) * v_i$$

11. Solve $A' * v = b'$ for $v = (f, t_z)$.

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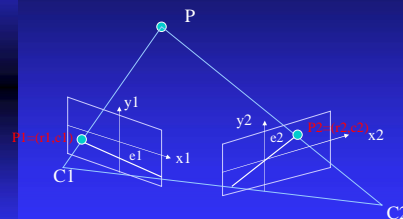
Almost there

12. If f is negative, change signs (see text).
13. Compute the lens distortion factor k and improve the estimates for f and t_z by solving a nonlinear system of equations by a nonlinear regression.
14. All parameters have been computed.

Use them in 3D data acquisition systems.

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We use them for general stereo.



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For a correspondence $(r1,c1)$ in image 1 to $(r2,c2)$ in image 2:

1. Both cameras were calibrated. Both camera matrices are then known. From the two camera equations we get

4 linear equations in 3 unknowns.

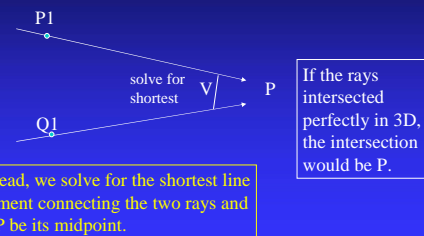
$$\begin{aligned} r1 &= (b11 - b31*r1) * x + (b12 - b32*r1) * y + (b13 - b33*r1) * z \\ c1 &= (b21 - b31*c1) * x + (b22 - b32*c1) * y + (b23 - b33*c1) * z \end{aligned}$$

$$\begin{aligned} r2 &= (c11 - c31*r2) * x + (c12 - c32*r2) * y + (c13 - c33*r2) * z \\ c2 &= (c21 - c31*c2) * x + (c22 - c32*c2) * y + (c23 - c33*c2) * z \end{aligned}$$

Direct solution uses 3 equations, won't give reliable results.

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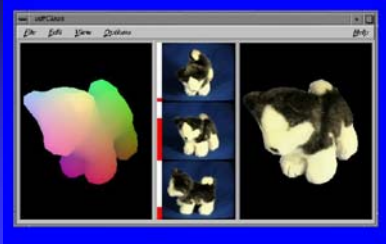
Solve by computing the closest approach of the two skew rays.



Instead, we solve for the shortest line segment connecting the two rays and let P be its midpoint.

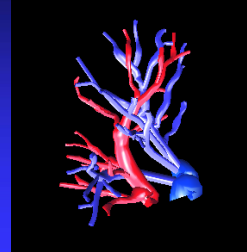
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Application: Kari Pulli's Reconstruction of 3D Objects from light-striping stereo.



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Application: Zhenrong Qian's 3D Blood Vessel Reconstruction from Visible Human Data



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