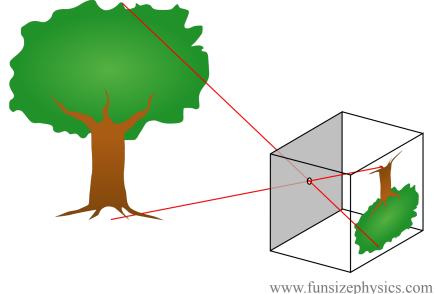
Perspective Correction and Motion Estimation

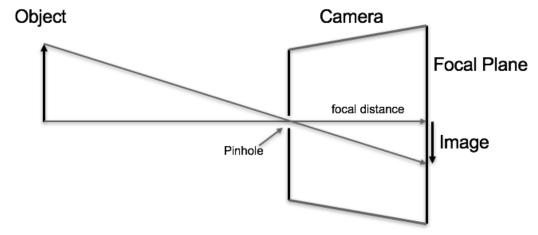
CU Boulder

The Pinhole Camera

- Image is inverted
- Infinite depth of field



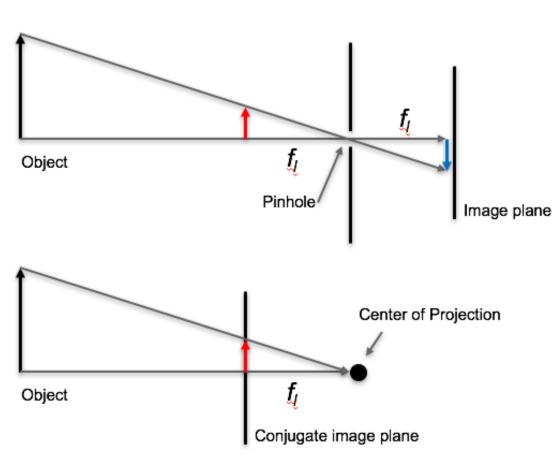
www.funsizephysics.con



Simplest Camera Model cont.

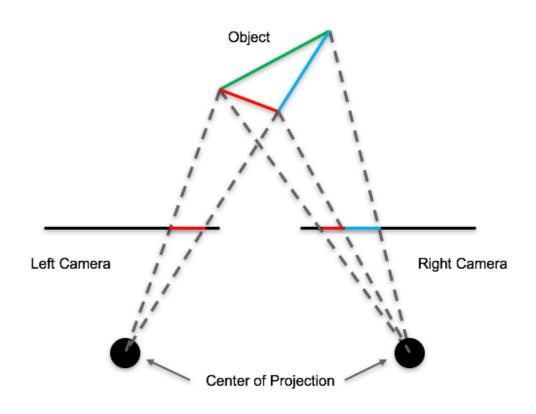
CU Boulder

- ▶ Upside down □ Upside up



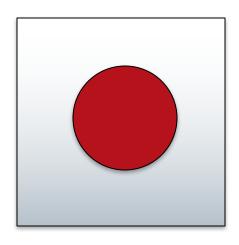
Stereo Camera Model

 Camera diagrams like this are commonly encountered in computer vision and graphics

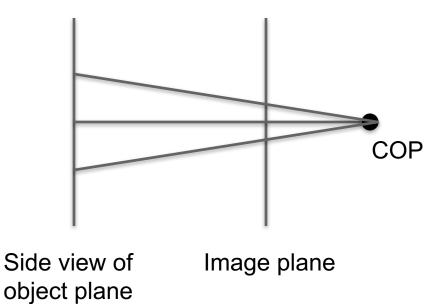


Projective Distortion

A circle in an object plane parallel to the image plane images as a circle



Front view of object plane

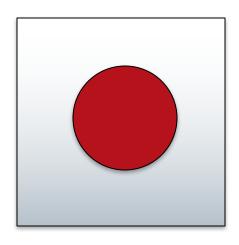


Front view of image plane picture

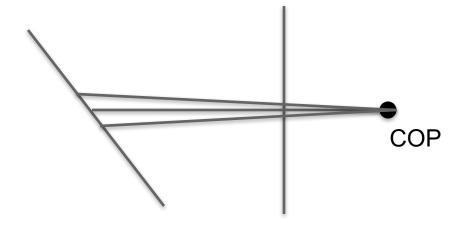


Perspective Distortion cont.

A circle in a plane tilted relative to the image plane images as an ellipse



Front view of object plane



Side view of object plane

Image plane

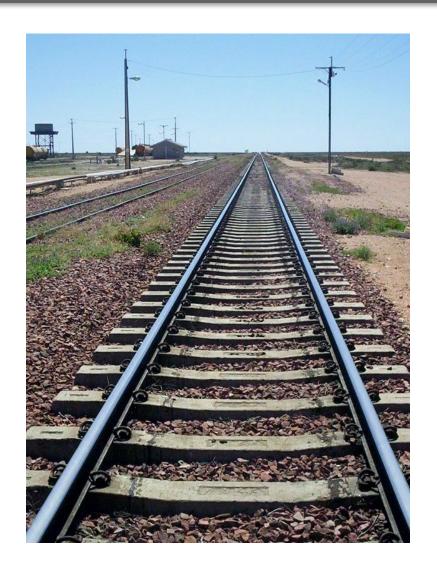
Front view of image plane picture



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Another example

- Parallel lines lying in a plane that is tilted relative to the image plane appear to converge at infinity
- If the lines were lying in a plane parallel to the image plane, they would appear parallel



Perspective Distortion cont.

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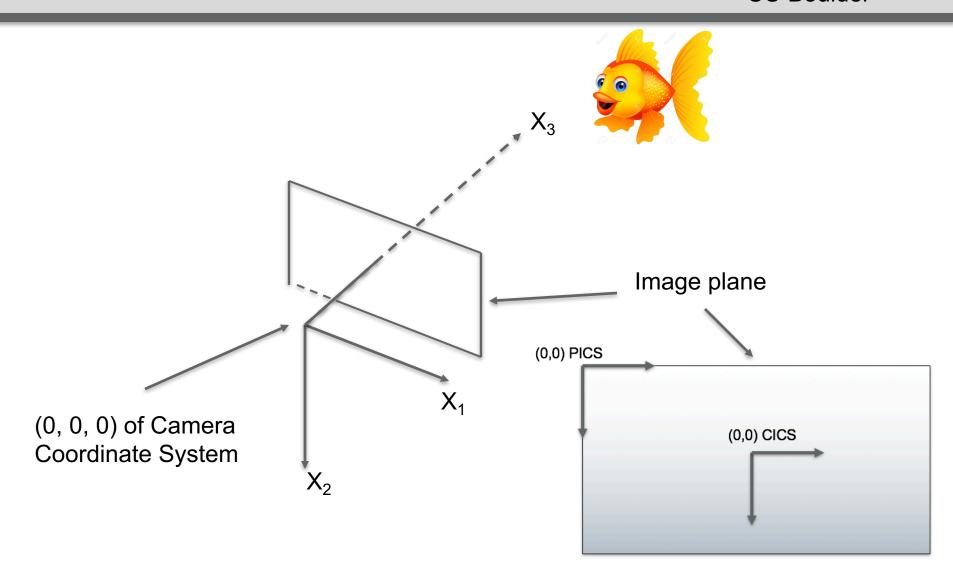
Distortions include

- Shape is not preserved
 - √ e.g. cirlces become ellipses
- Distance between two points is not preserved
 - ✓ e.g. parallel lines converge
- Angles are not preserved
 - ✓ e.g. Imagine rotating a right triangle lying in a plane that starts parallel to the image plane and ends perpendicular to it



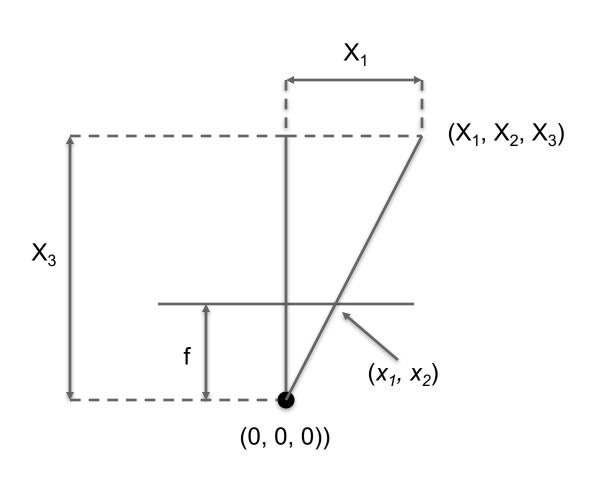
Image Coordinate Systems

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PICS: Pixel Image Coordinate System CICS: Canonical Image Coordinate System

Projection Diagram



This diagram is in the X_1X_3 plane.

Visualize the point (X_1, X_2, X_3) as lying above the plane so that $X_2 < 0$

$$\frac{f}{X_3} = \frac{x_1}{X_1}$$

$$x_1 = f \frac{X_1}{X_3}$$

Non-linear Projection Formulas

Therefore, the Camera Coordinates (X₁, X₂, X₃) are related to the canonical image coordinates (x₁, x₂) via

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} f & 0 \\ 0 & f \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \end{bmatrix} \frac{1}{X_3}$$

This term is UGLY and makes the relationship between (X_1, X_2, X_3) and (x_1, x_2) non-linear. There is a better way!

Linearizing the Projection Formulas

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Consider the following <u>linear</u> equation

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} f & 0 & 0 & 0 \\ 0 & f & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \\ 1 \end{bmatrix}$$

- We put a "1" under the (X_1, X_2, X_3) vector
- We can get the CIC coordinates of the projection point, \underline{x} , by computing x_1/x_3 and x_2/x_3

This equation uses <u>homogenous</u> coordinates

Homogenous Coordinates

- In Euclidean geometry, a point in N-D space is described by an N-D vector
- In Projective geometry, a point in N-D space is described by an (N+1)-D vector (homogenous coordinates)
- The last coordinate is a <u>multiplier</u> of the first N coordinates: let its value be k
- A point now has <u>multiple representations</u>, one for each value of k
 - Example: the 2-D Euclidean point, (rows, cols), is the SAME POINT as <u>ANY</u> 3-D vector of the form

(k*rows, k*cols, k)

Homogenous Coordinates Rules

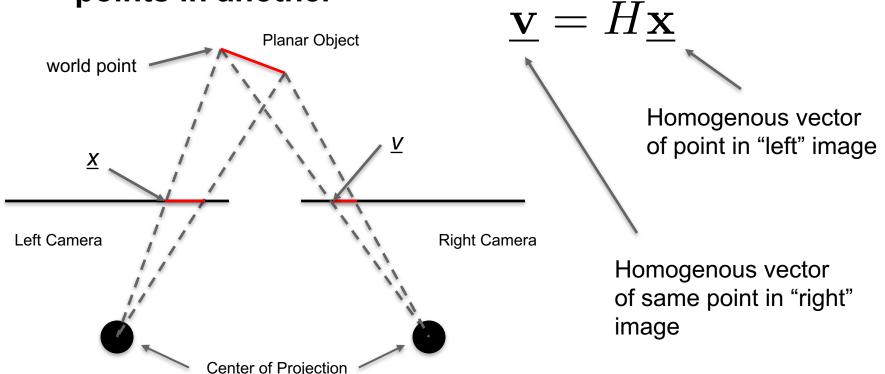
We refer to such a representation as a homogenous vector

$$(k*rows, k*cols, k) \leftarrow$$
 homogenous vector

- We say we are using <u>homogenous coordinates</u>
- We can find the Euclidean point corresponding to a homogenous vector by dividing the first N components by the last component
 - 2-D Example: (rows, cols) = (k*rows/k, k*cols/k)(12, 8, 2) \rightarrow (6, 4)

Projective Geometry Fact

Given a set of world points known to lie on a <u>plane</u>, there exists a matrix, H, that maps the (row, col) image points in one camera to the (row, col) image points in another



Take a course in computer vision for the details!

We seek a matrix H that maps lines converging at infinity to parallel lines in the Euclidean plane

$$\underline{\mathbf{v}} = H\underline{\mathbf{x}}$$

Homogenous vector in processed image



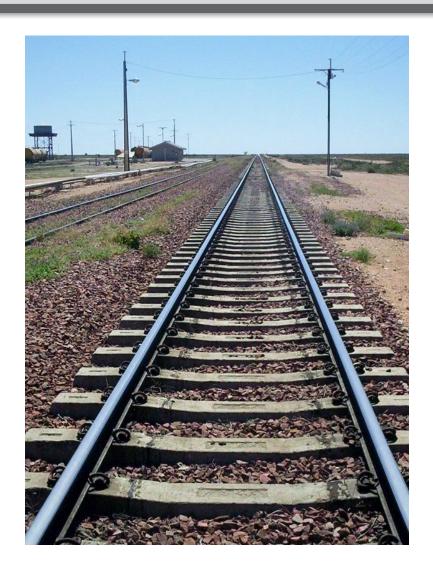
Homogenous vector in "distorted" image

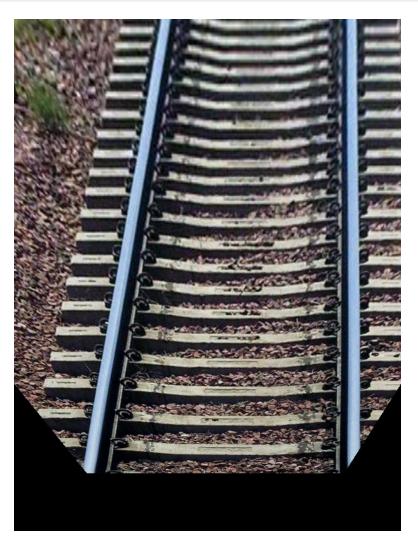


Note that the points on the wall lie in a plane

Perspective Distortion cont.

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Perspective Distortion and its removal via image processing

Finding the Matrix *H*

- Assume we have specified N corresponding points in both images
 - We are free to choose points in one image and specify where they want them to be after mapping
 - For example, assume below are 4 of the N point correspondences we have specified

$$(c_1, d_1) \to (a_1, b_1)$$

 $(c_2, d_2) \to (a_2, b_2)$
 $(c_3, d_3) \to (a_3, b_3)$
 $(c_4, d_4) \to (a_4, b_4)$

(c, d) pairs are in the "distorted" image. (a, b) pairs are in the processed image. Coordinates are in PICS

The equations we need are these

$$\begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & 1 \end{bmatrix} \begin{bmatrix} c \\ d \\ 1 \end{bmatrix} \text{ and } \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} v_1/v_3 \\ v_2/v_3 \end{bmatrix}$$

- We need to solve for the matrix values h_{ij}
 - There are 8 unknown variables in H

Note: you might wonder why we can specify the (3, 3) component of H to be one. The reason: since we are using homogenous coordinates, multiplying both sides of the equation by a constant has no net effect. So whatever h_{33} is, we assume we have multiplied the equation by $1/h_{33}$.

• By solving for v_1/v_3 and v_2/v_3 the previous equations yield

$$a = \frac{h_{11}c + h_{12}d + h_{13}}{h_{31}c + h_{32}d + 1}$$
$$b = \frac{h_{21}c + h_{22}d + h_{23}}{h_{31}c + h_{32}d + 1}$$

Each pair of corresponding points yields two equations, so we need a minimum of four corresponding points to solve for the eight variables in H. In practice, we want more.

These two equations can be rewritten in matrix form

$$\begin{bmatrix} c & d & 1 & 0 & 0 & 0 & -ac & -ad \\ 0 & 0 & 0 & c & d & 1 & -bc & -bd \end{bmatrix} \begin{bmatrix} h_{11} \\ h_{12} \\ h_{13} \\ h_{21} \\ h_{22} \\ h_{23} \\ h_{31} \\ h_{32} \end{bmatrix} = \begin{bmatrix} a \\ b \end{bmatrix}$$

Each new pair of corresponding points adds 2 more rows to the matrix and 2 more rows to the right-hand side vector

$$\begin{bmatrix} c & d & 1 & 0 & 0 & 0 & -ac & -ad \\ 0 & 0 & 0 & c & d & 1 & -bc & -bd \end{bmatrix} \begin{bmatrix} h_{11} \\ h_{12} \\ h_{13} \\ h_{21} \\ h_{22} \\ h_{23} \\ h_{31} \\ h_{32} \end{bmatrix} = \begin{bmatrix} a \\ b \end{bmatrix}$$

These grow downward with more corresponding points

▶ This matrix equation can be solved for <u>h</u>

- If more than 4 points are used, solving for H becomes a linear regression problem
 - ✓ Use the pseudo inverse
- <u>h</u> can then be reshaped into the 3 x 3 matrix H

For this lab we will code up the pseudo-inverse calculation ourselves (you can <u>check</u> your result with the library function if you wish)

Removing Perspective Distortion

- Constructing the corrected image, once you have H, proceeds as follows:
 - Step through the pixels in the corrected image using for() loops
 - ✓ Declare an array of all black pixels as the corrected image
 - Use H⁻¹ to find corresponding pixels in the input image
 - ✓ The values you find will not be integers
 - ✓ If the values are outside the edge of the picture, skip to the next pixel
 - Bi-linearly interpolate values in the input image
 - ✓ Code to do this is provided in the lab writeup
 - Assign the interpolated result to the corrected image pixel

Measuring Points in an Image

- You can do this with photoshop, GIMP, or other offthe-shelf image display tools
- I have also uploaded to the lab 3 module a script called get_mouse_pos.py
 - It requires the module "pyuserinput" to run
 - You can install it via: pip3 install pyuserinput
- To use get_mouse_pos.py
 - Display the picture
 - Run the script in a terminal window
 - Click the upper left-hand corner of the picture (make sure the picture is displayed at its full resolution)
 - Click on the image points whose coordinates you desire (right-click to exit)