

# Spherical Warping

Image warping



we did this using homographies

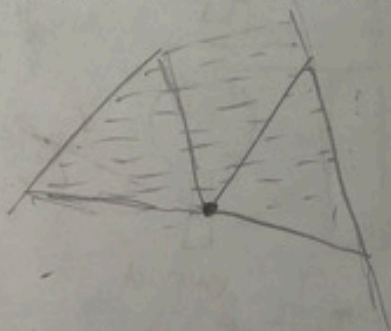


We can view homographies as linear transformations if we consider that our planes are in 3D



project one image into another image's view plane

However, we cannot handle cases like this;



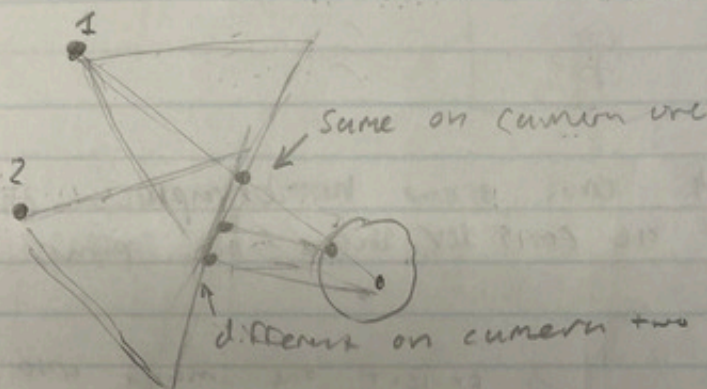
So, let's project onto a sphere



Nice and separate!

Going from  $(x, y)$  to  $(\phi, \theta)$  space.

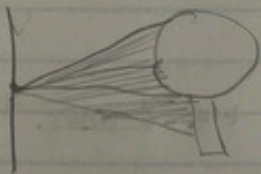
Question: These methods hinge on a single viewpoint. what if we have multiple?



This is bad for panorama stitching, but great for depth estimation!



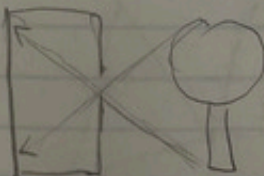
576 cam Mk II:



Too many locations contribute to one point on the sensor



576 cam Mk II:



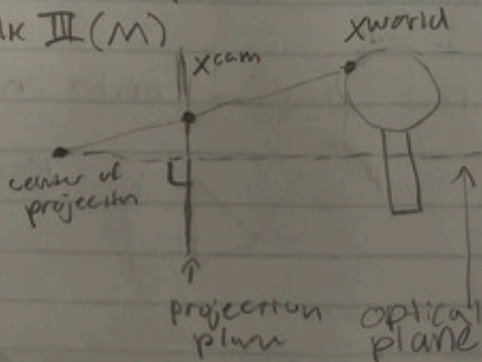
Good! Realistic

MATH!



576 cam Mk III (M)

$x_{img}$

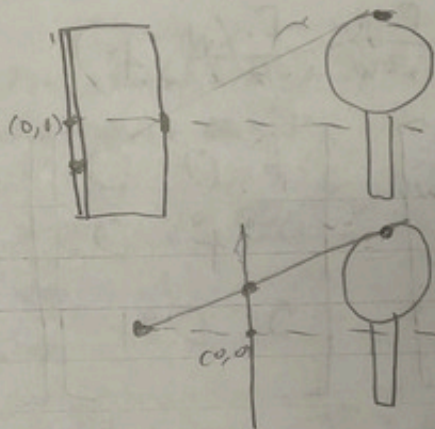


Mathematically useful to us...

How do we get  $X^{img}$  from  $X^{world}$ ?

$$X^{img} = \begin{bmatrix} \text{Intrinsics} \\ \text{focal length,} \\ \text{pixel coord. system} \end{bmatrix} \begin{bmatrix} \text{Projection} \\ \text{3D to 2D} \end{bmatrix} \begin{bmatrix} \text{Extrinsics} \\ \text{camera pose} \\ X^{cam} \text{ (in world)} \end{bmatrix} X^{world}$$

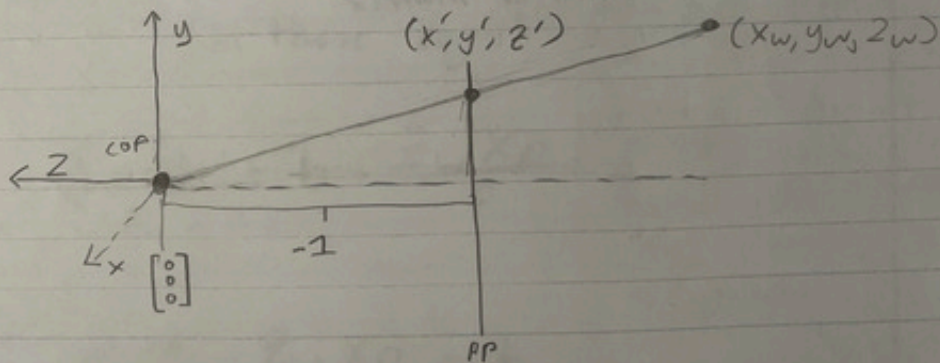
Activity: HM Problem 1



$y$  is flipped

$$(x^{MkIII}, y^{MkIII}) = (-x^{MkI}, -y^{MkI})$$

HM Problem 2



Find  $(x', y', z')$ !

Homogeneous looking!

$$(x', y', z') = \left( \frac{x_w}{-z_w}, \frac{y_w}{-z_w}, -1 \right)$$

Lets express this using linear algebra

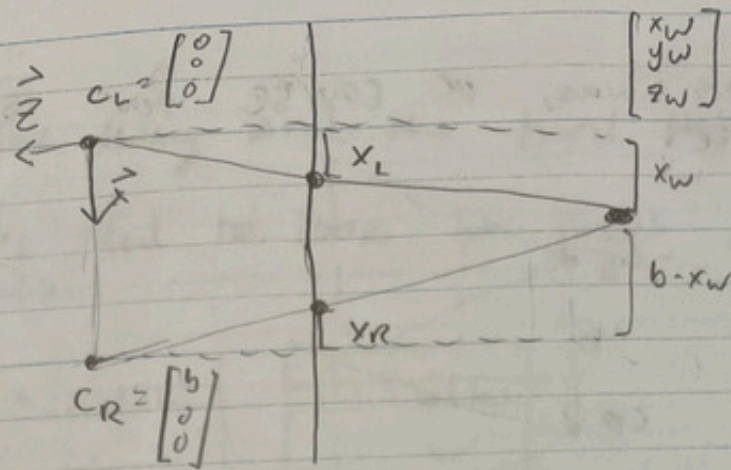
$$\begin{bmatrix} \frac{x_w}{z_w} \\ \frac{y_w}{z_w} \\ 1 \end{bmatrix} \sim \begin{bmatrix} x_w \\ y_w \\ z_w \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_w \\ y_w \\ z_w \end{bmatrix}$$

True! But not at all useful... Boring!

$$(x', y', z') = \left( -\frac{f \cdot y_w}{z_w}, -\frac{f \cdot x_w}{z_w}, f \right)$$

$$\begin{bmatrix} \frac{f x_w}{z_w} \\ \frac{f y_w}{z_w} \\ 1 \end{bmatrix} \sim \begin{bmatrix} f x_w \\ f y_w \\ z_w \end{bmatrix} = \begin{bmatrix} -f & 0 & 0 \\ 0 & -f & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_w \\ y_w \\ z_w \end{bmatrix}$$

Better! Version 2.1 of our projection is  
Extrinsic matrix



Assume (for now)  $x_w$  is known (we don't, but things will work out)

$$\frac{z_w}{f} = \frac{x_w}{x_L}, \quad \frac{b - x_w}{x_R} = \frac{z_w}{f}$$

$$x_w = \frac{z_w x_L}{f} \quad x_w = b - \frac{z_w x_R}{f}$$

Set these equal! Goodbye  $x_w$ !

$$\frac{z_w x_L}{f} = b + \frac{z_w x_R}{f}$$

$$\frac{z_w x_L}{f} - \frac{z_w x_R}{f} = b$$

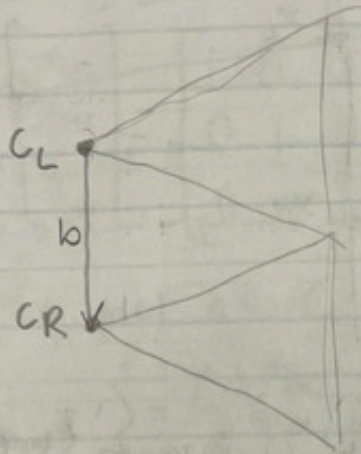
$$z_w \left( \frac{x_L - x_R}{f} \right) = b$$

$$z_w = \frac{fb}{x_L - x_R}$$

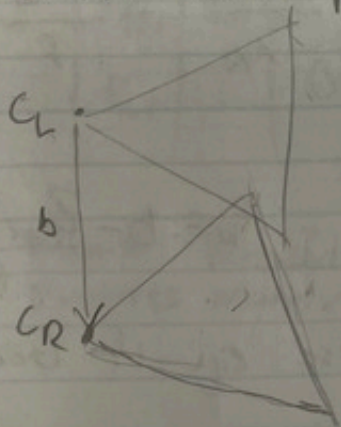
Aside!

Depth is inversely proportional to disparity.

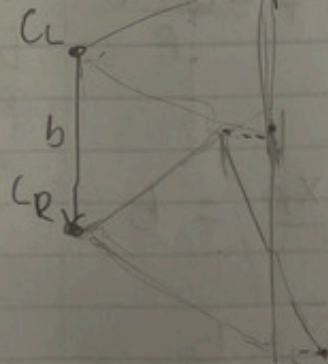
This was, of course for the simple case:



But what if we have cameras with different view planes?



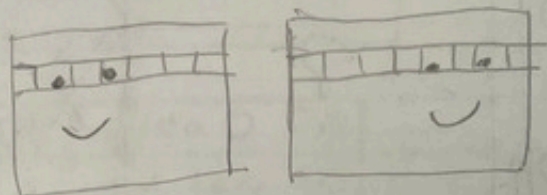
We could do some painful 3D math, or...



We can project  $C_R$ 's view plane onto  $C_L$ 's.

The nice thing and the hard thing:

We only need to look at rows to find matches:



Less work (good!)

But more chances to make mistakes (not good!)

So we need some ability to find good matches...

The correspondence problem:

We need a way to assign a cost to some matching between two rows:

SSD - Sum. Squared Difference

SAD - Sum. Abs. Difference

CC - cross correlation...

If we have some pattern we want to find, if we use it as a filter in cross correlation it should create "bright spots" (large numbers) in areas that are similar to it.

However, we want these detections to be consistent and comparable from one another... so we normalized.

# Camera matrix

From last time!

$$\begin{array}{c} \text{Intrinsic} \\ \begin{bmatrix} f & 0 & c_x & 0 \\ 0 & f & c_y & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \end{array} \begin{array}{c} \text{Extrinsic} \\ \begin{bmatrix} ? \\ ? \\ ? \\ ? \end{bmatrix} \end{array} \begin{array}{c} \vec{X}_{\text{world}} \leftarrow \\ \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} \end{array}$$

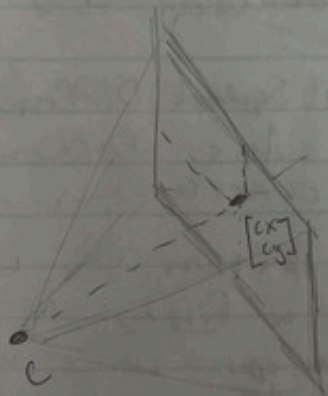
$3 \times 4$        $4 \times 4$

$X_{\text{cam}}$

Why the extra column? 3D Homogeneous coordinates!

$$\begin{bmatrix} c_x \\ c_y \end{bmatrix}$$

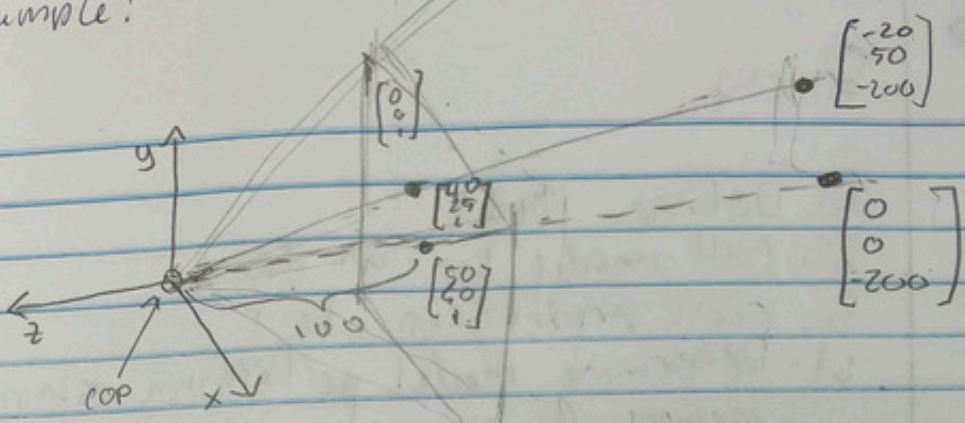
The principal point: The location of the optical axis in pixel coordinates.



Most of the time, this is just the center of the image  $(\frac{w}{2}, \frac{h}{2})$



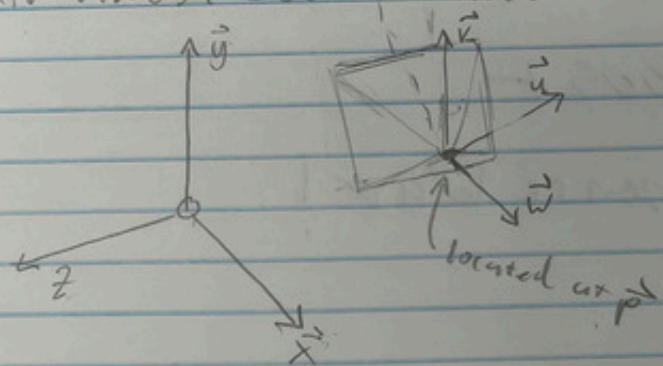
Example:



Intrinsic:

$$\begin{bmatrix} 100 & 0 & 50 \\ 0 & 100 & 50 \\ 0 & 0 & 1 \end{bmatrix}$$

? Extrinsic: Our camera isn't nicely placed:



We want to be able to work relative to the camera... so we want  $M$  s.t.

$$\begin{bmatrix} M \\ x \\ y \\ z \end{bmatrix} = \begin{bmatrix} u \\ v \\ w \end{bmatrix}$$

$$\begin{bmatrix} | & | & | & | \\ \vec{u} & \vec{v} & \vec{w} & \vec{p} \\ | & | & | & | \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} u \\ v \\ w \\ 1 \end{bmatrix} \approx \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} \quad \text{So, } M = \begin{bmatrix} | & | & | & | \\ \vec{u} & \vec{v} & \vec{w} & \vec{p} \\ | & | & | & | \\ 0 & 0 & 0 & 1 \end{bmatrix}^{-1}$$