Lecture 11: Camera Models

COMP 590/776: Computer Vision

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TA: Mykhailo (Misha) Shvets



Course Website: Scan Me!

Breaking out of 2D

...now we are ready to break out of 2D







And enter the real world!



NeRF in the wild (will get to it towards the end!)



NeRF in the Wild, Martin-Brualla, Radwan et al. CVPR 2021

Nerfstudio: Colab friendly github repo for NeRF (some of you might wanna try this for your project!)



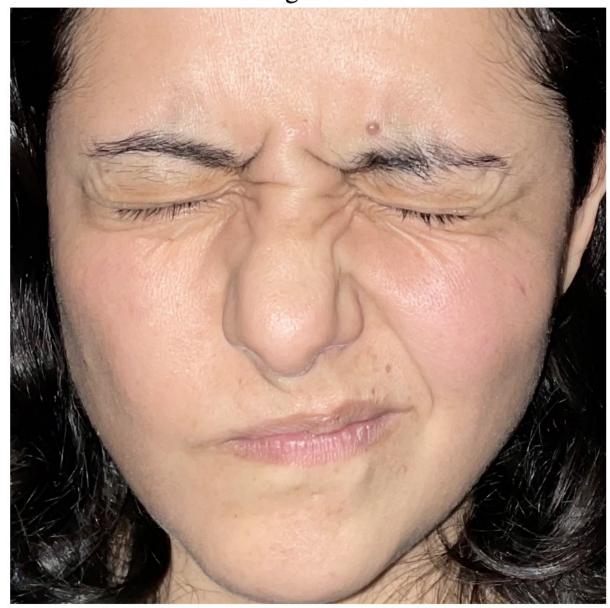


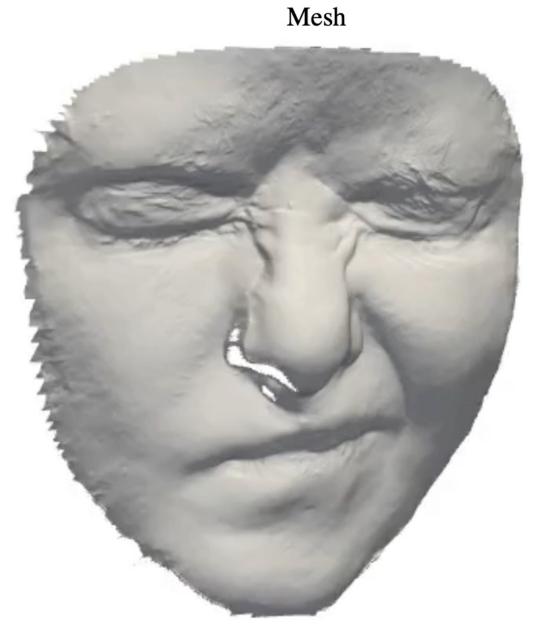
Captured Images: Right



Single iPhone Image with Built-In Flash

Image 1/1





Lectures in 3D Vision

- Fundamental Concepts (4 lectures)
 - Modeling camera and 3D->2D projection (2 lectures)
 - 2-view geometry & Stereo Vision (2 lectures)
- 3D Reconstruction techniques (2 lectures)
 - Multiview Stereo (MVS)
 - Structure from Motion (SfM) + SLAM
 - Photometric Stereo (PS)
- Deep Learning + 3D Vision (2 lectures)
 - Deep Learning + MVS, SfM, PS
 - Neural Radiance Fields (NeRFs)

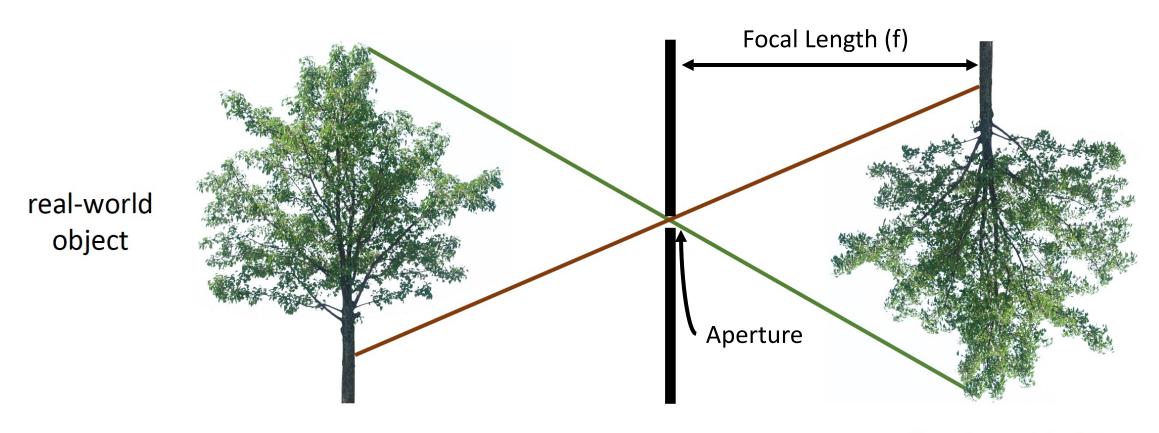
Today's Class

- Pinhole & Lens Camera
- Camera Parameters
 - Extrinsic
 - Intrinsic
- Perspective Distortion
- Other Projection models
- How to calibrate camera, i.e. estimate camera parameters (next class)

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Pinhole Camera



copy of real-world object (inverted and scaled)

Focal length

• Can think of as "zoom"



24mm



50mm

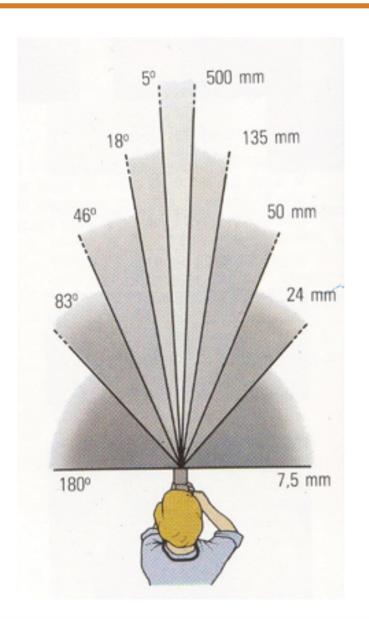


200mm



Also related to field of view (inversely)

Focal length in practice



24mm



50mm

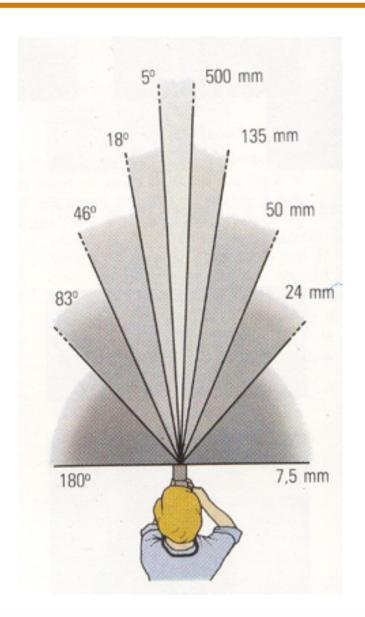


135mm



Fredo Durand

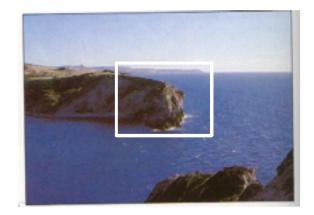
Focal length = cropping



24mm



50mm

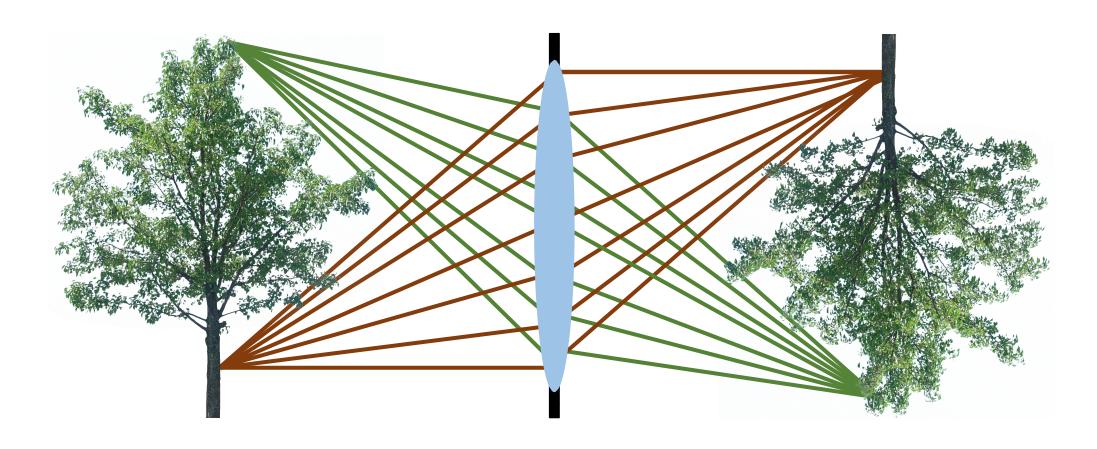


135mm



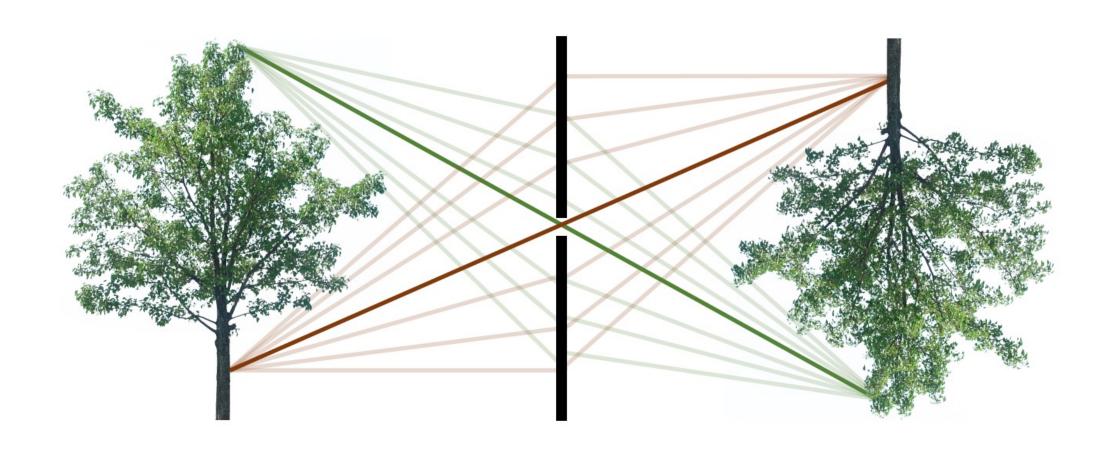
Fredo Durand

The lens camera



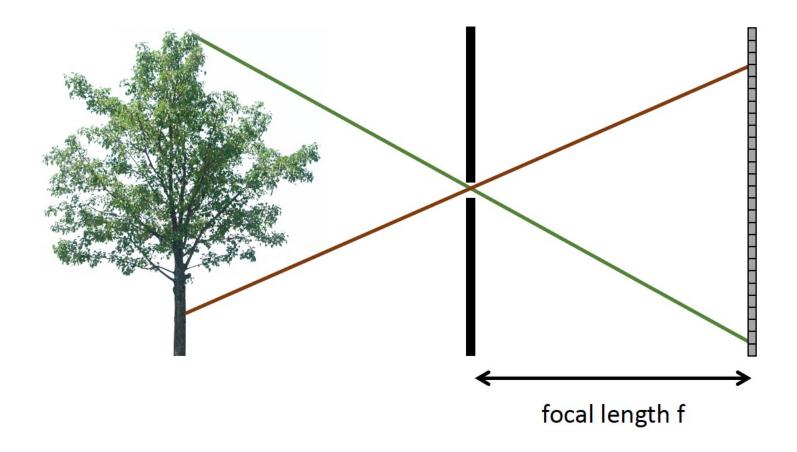
Lenses map "bundles" of rays from points on the scene to the sensor.

The pinhole camera



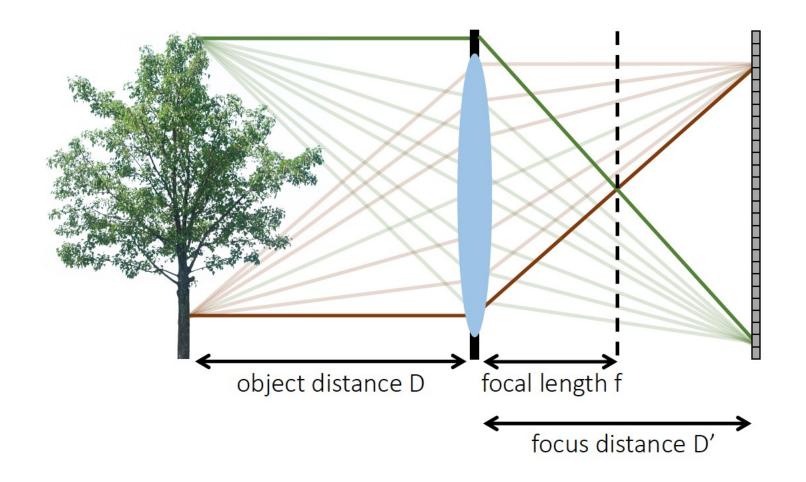
Central rays propagate in the same way for both models!

Important Difference: focal length



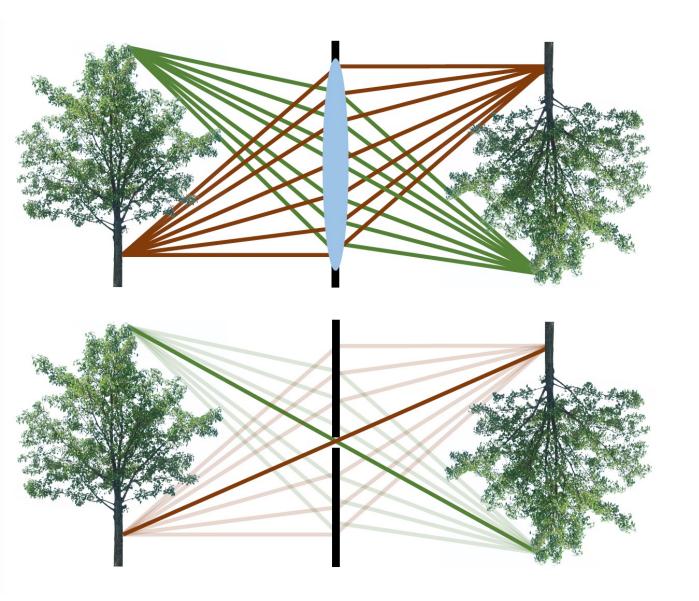
In a pinhole camera, focal length is distance between aperture and sensor

Important Difference: focal length



In a lens camera, focal length is distance where parallel rays intersect

Describing both lens and pinhole cameras



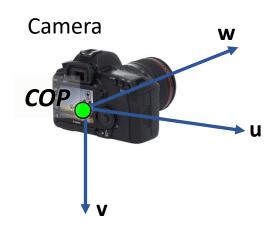
We can derive properties and descriptions that hold for both camera models if:

- We use only central rays.
- We assume the lens camera is in focus.
- We assume that the focus distance of the lens camera is equal to the focal length of the pinhole camera.

From now on, we will describe cameras as pinhole cameras!
Focal length will mean aperture-sensor distance.

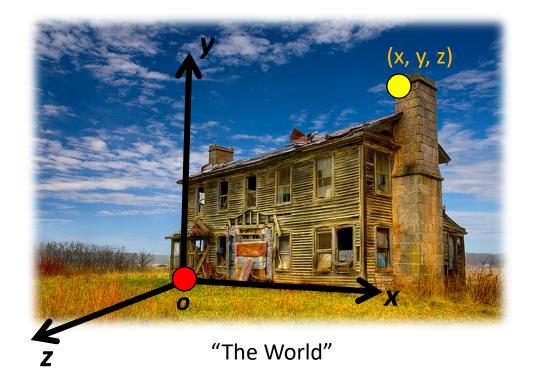
Camera parameters

How can we model the geometry of a camera?



Three important coordinate systems:

- 1. World coordinates
- 2. Camera coordinates
- 3. Image coordinates



How do we project a given world point (x, y, z) to an image point?

Coordinate frames

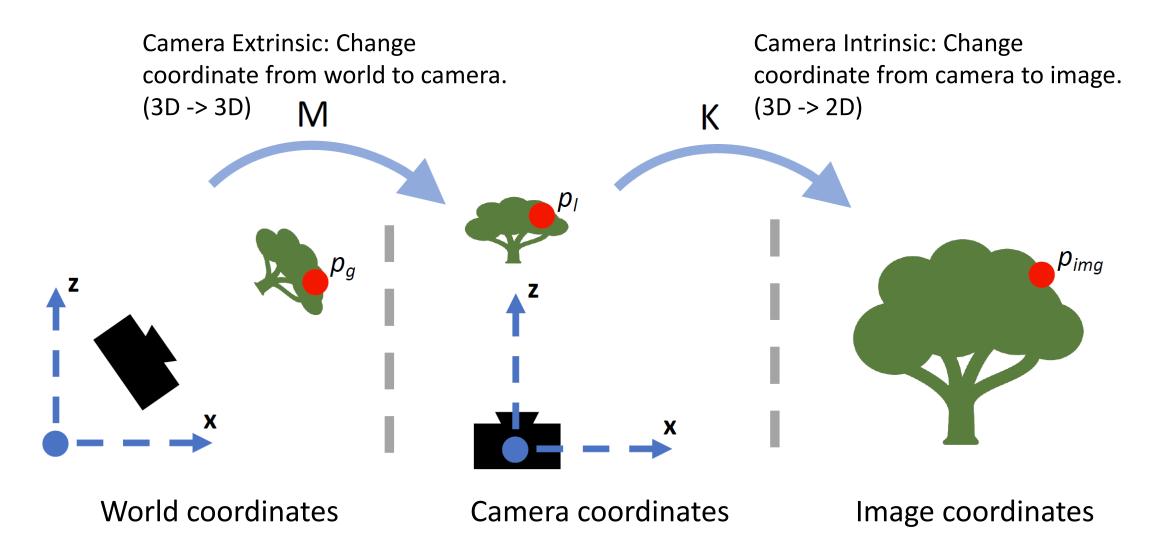


Figure credit: Peter Hedman

Today's Class

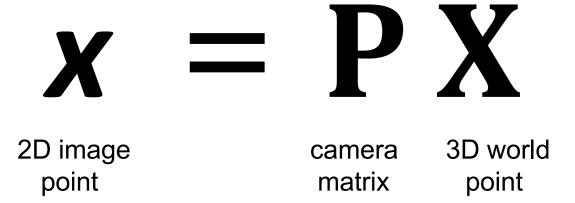
- Pinhole & Lens Camera
- Camera Parameters
 - Extrinsic
 - Intrinsic
- Perspective Distortion
- Other Projection models
- How to calibrate camera, i.e. estimate camera parameters (next class)

Camera parameters

To project a point (x, y, z) in world coordinates into a camera

- First transform (x, y, z) into camera coordinates
- Need to know
 - Camera position (in world coordinates)
 - Camera orientation (in world coordinates)
 - Together they form *Camera Extrinsics*
- Then project into the image plane to get image (pixel) coordinates
 - Need to know Camera Intrinsics

A camera is a mapping between the 3D world and a 2D image



x = PX

$$\begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = \begin{bmatrix} p_1 & p_2 & p_3 & p_4 \\ p_5 & p_6 & p_7 & p_8 \\ p_9 & p_{10} & p_{11} & p_{12} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

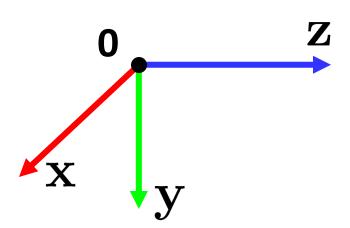
homogeneous image 3 x 1 Camera matrix 3 x 4 homogeneous world point 4 x 1

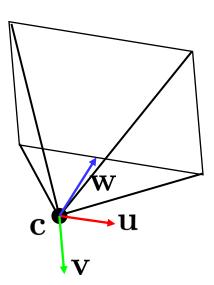
- How do we get the camera to "canonical form"?
 - Canonical form: Center of projection at the origin, x-axis points right, y-axis points down, z-axis points forwards

 X_w -> location of a point in world coordinate.

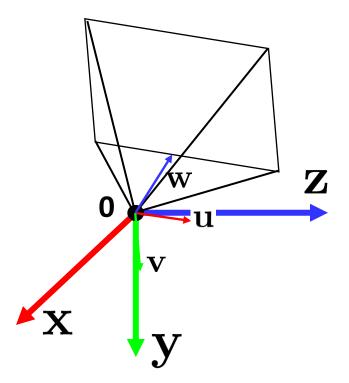
 X_c -> location of a point in camera coordinate.

Step 1: Translate by -c



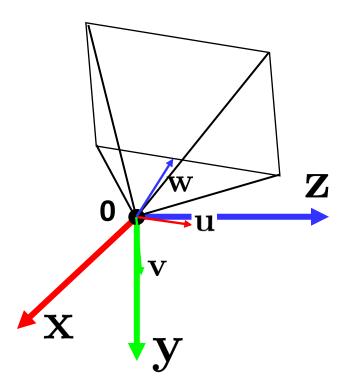


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Step 1: Translate by -c

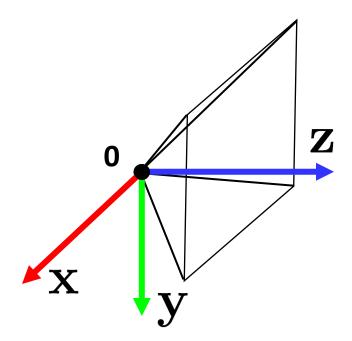
- How do we get the camera to "canonical form"?
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Step 1: Translate by -c

Step 2: Rotate by R

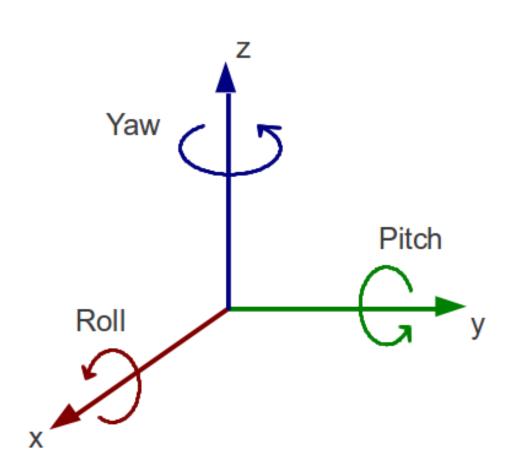
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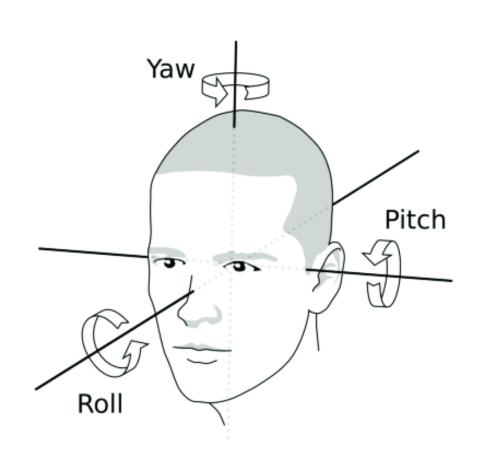


Step 1: Translate by -c

Step 2: Rotate by R

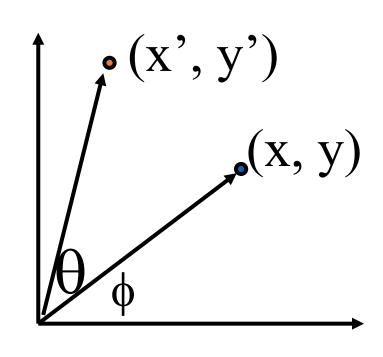
How do we represent 3D rotation?





Euler Angles

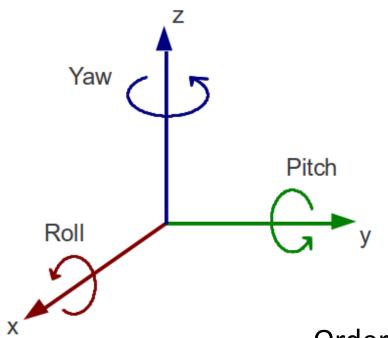
What did we do with 2D rotation?



$$\mathbf{R} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$
R

Rotation matrix in 3D



$$R_x(heta) = egin{bmatrix} 1 & 0 & 0 \ 0 & \cos heta & -\sin heta \ 0 & \sin heta & \cos heta \end{bmatrix}$$

$$R_y(heta) = egin{bmatrix} \cos heta & 0 & \sin heta \ 0 & 1 & 0 \ -\sin heta & 0 & \cos heta \end{bmatrix}$$

$$R_z(heta) = egin{bmatrix} \cos heta & -\sin heta & 0 \ \sin heta & \cos heta & 0 \ 0 & 0 & 1 \end{bmatrix}$$

Order of applying rotation matters (composition of 3D rotations is not commutative.)

$$R = R_z(lpha)\,R_y(eta)\,R_x(\gamma) = egin{bmatrix} \coslpha & -\sinlpha & 0 \ \sinlpha & \coslpha & 0 \ 0 & 0 & 1 \end{bmatrix} egin{bmatrix} \coseta & 0 & \sineta \ 0 & 1 & 0 \ -\sineta & 0 & \coseta \end{bmatrix} egin{bmatrix} 1 & 0 & 0 \ 0 & \cos\gamma & -\sin\gamma \ 0 & \sin\gamma & \cos\gamma \end{bmatrix}$$

How to derive camera extrinsics? [*]

Show on board.

M = R [I | -C] (translate first then rotate)

or

M = [R | t], where t=-RC (rotate first then translate)

Coordinate frames

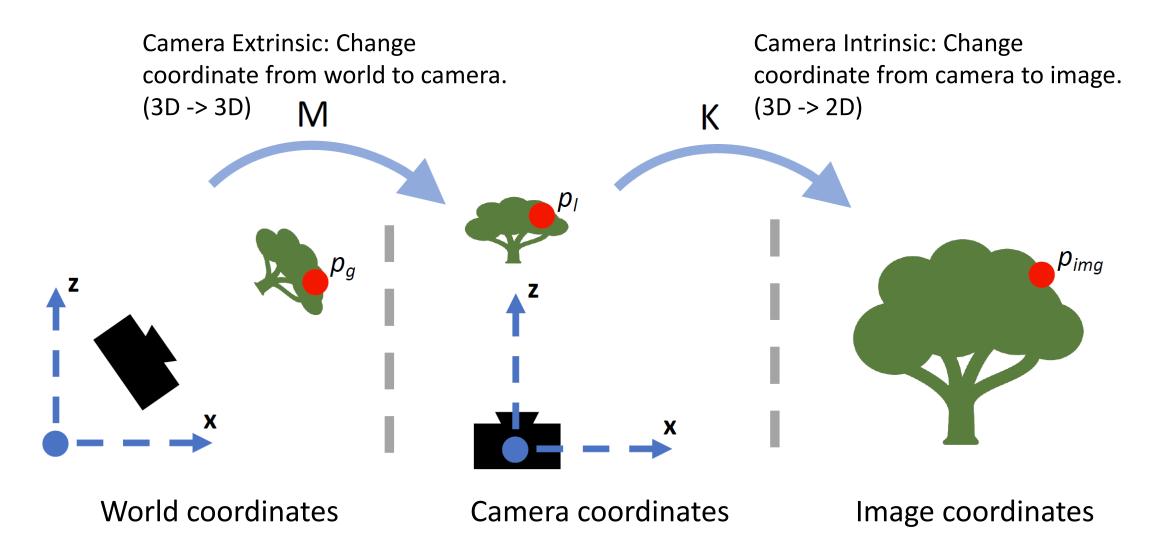
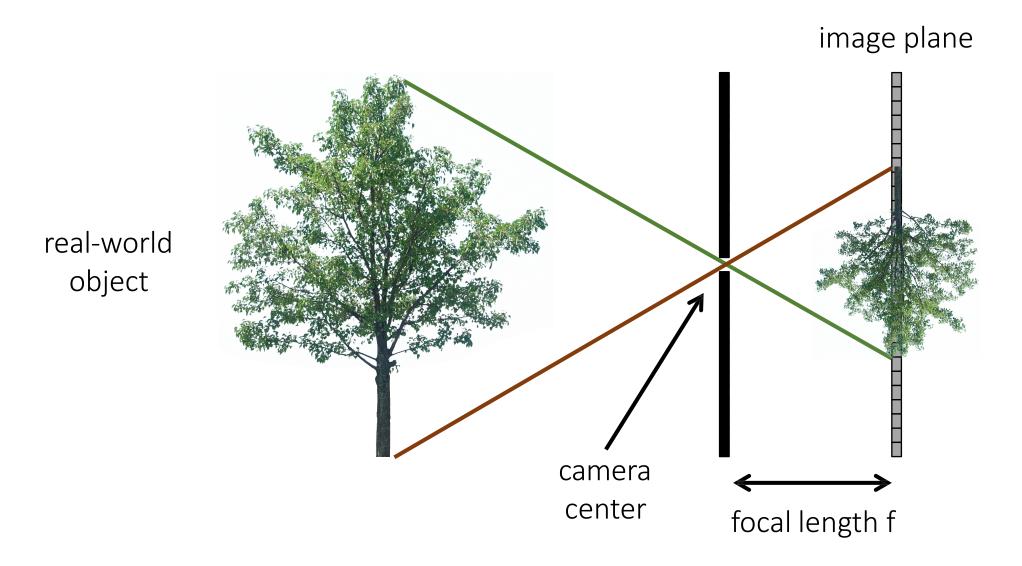


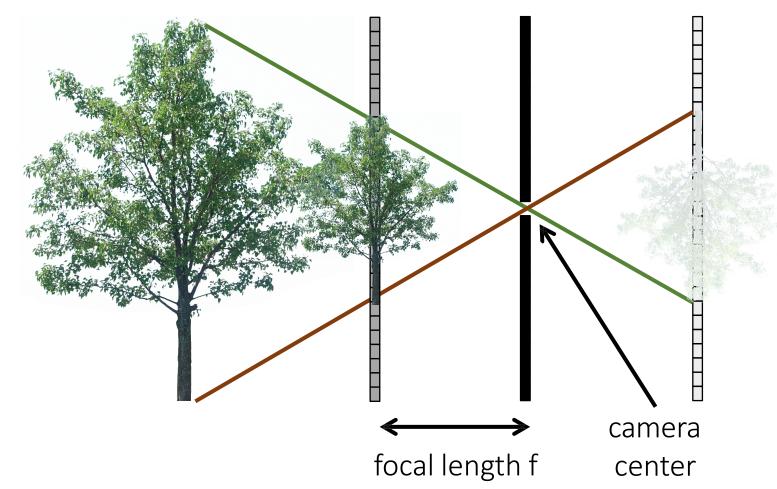
Figure credit: Peter Hedman

The pinhole camera



The (rearranged) pinhole camera

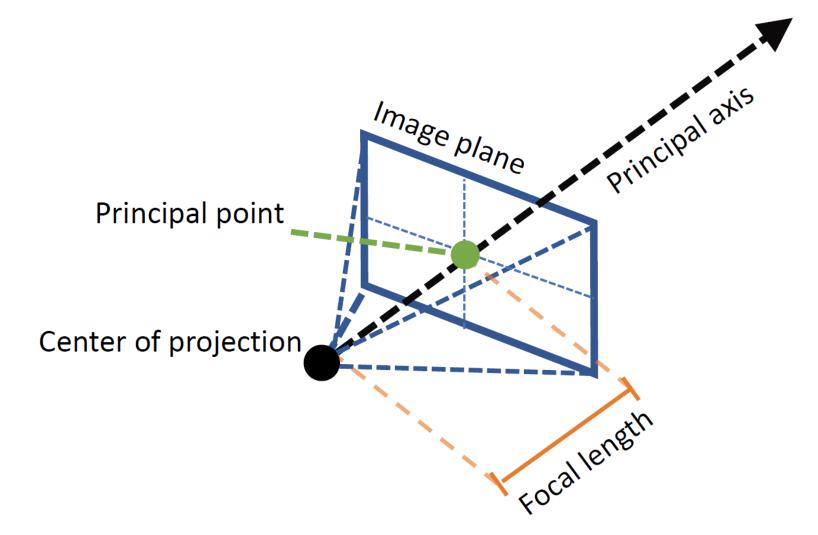




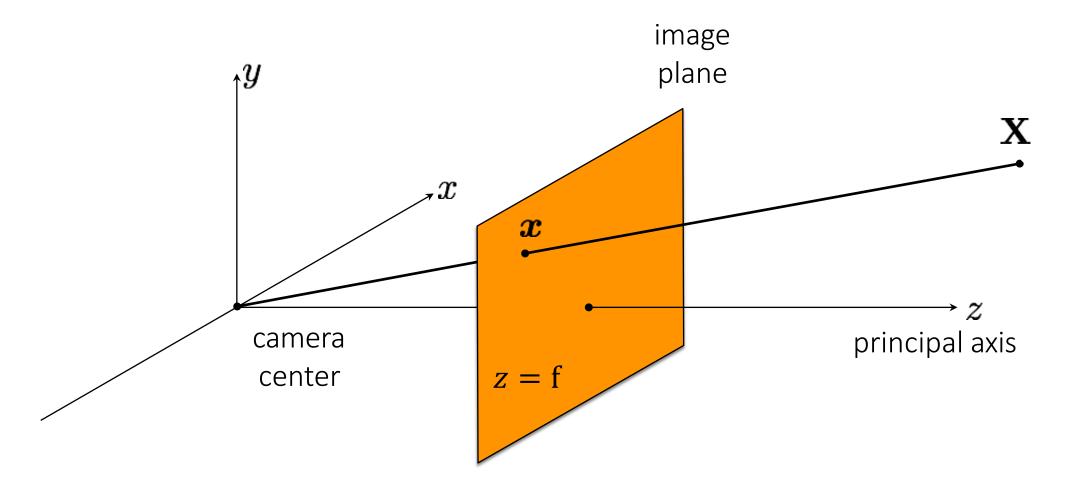
real-world

object

Geometric Model: A Pinhole Camera

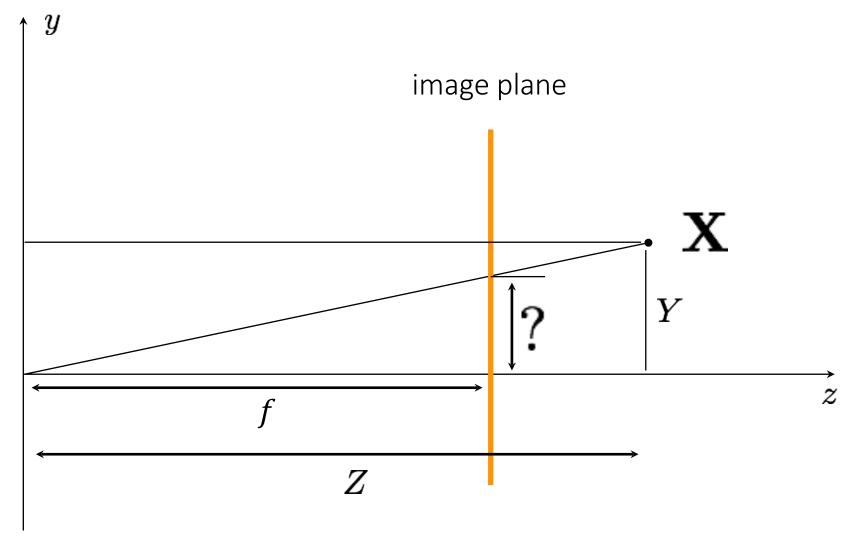


The (rearranged) pinhole camera



What is the equation for image coordinate x in terms of X?

The 2D view of the (rearranged) pinhole camera



What is the equation for image coordinate x in terms of X?

The pinhole camera matrix for arbitrary focal length

Relationship from similar triangles:

$$[X \quad Y \quad Z]^{\top} \mapsto [fX/Z \quad fY/Z]^{\top}$$

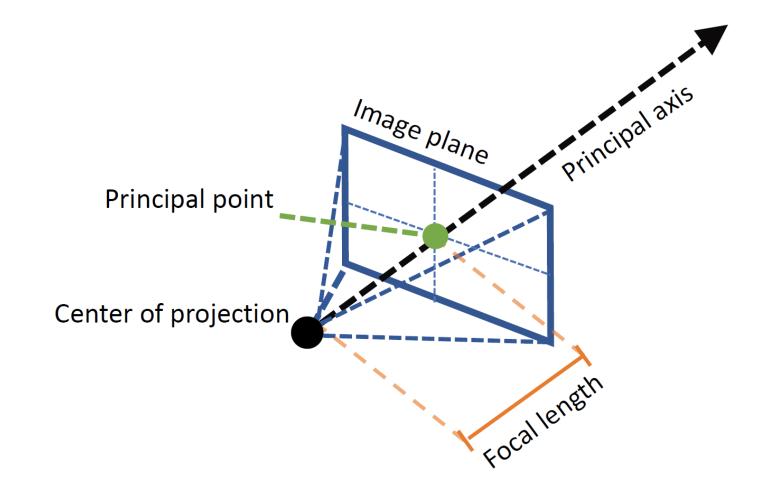
General camera model in homogeneous coordinates:

$$egin{bmatrix} \mathcal{X} \ \mathcal{Y} \ Z \end{bmatrix} &= \left[egin{array}{cccc} p_1 & p_2 & p_3 & p_4 \ p_5 & p_6 & p_7 & p_8 \ p_9 & p_{10} & p_{11} & p_{12} \end{array}
ight] \left[egin{array}{c} \mathcal{X} \ \mathcal{Y} \ Z \ 1 \end{array}
ight]$$

What does the pinhole camera projection look like?

$$\mathbf{P} = \left[egin{array}{cccc} f & 0 & 0 & 0 \ 0 & f & 0 & 0 \ 0 & 0 & 1 & 0 \end{array}
ight]$$

Geometric Model: A Pinhole Camera



Generalizing the camera matrix

In particular, the camera origin and image origin may be different:

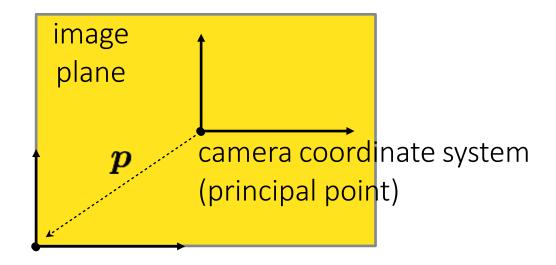


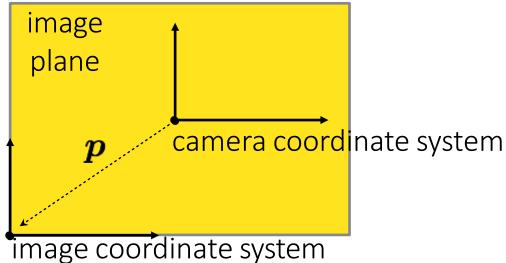
image coordinate system

How does the camera matrix change?

$$\mathbf{P} = \left[egin{array}{cccc} f & 0 & 0 & 0 \ 0 & f & 0 & 0 \ 0 & 0 & 1 & 0 \end{array}
ight]$$

Generalizing the camera matrix

In particular, the camera origin and image origin may be different:



How does the camera matrix change?

$$\mathbf{P} = \left[egin{array}{cccc} f & 0 & p_x & 0 \ 0 & f & p_y & 0 \ 0 & 0 & 1 & 0 \ \end{array}
ight]$$

shift vector transforming camera origin to image origin

Typical Intrinsics matrix

$$\mathbf{K} = \begin{bmatrix} f & 0 & c_x \\ 0 & f & c_y \\ 0 & 0 & 1 \end{bmatrix}$$

General case

$$\mathbf{K} = egin{bmatrix} f & 0 & c_x \ 0 & f & c_y \ 0 & 0 & 1 \end{bmatrix} \qquad \mathbf{K} = egin{bmatrix} f & s & c_x \ 0 & lpha f & c_y \ 0 & 0 & 1 \end{bmatrix}$$

2D affine transform corresponding to a scale by f (focal length) and a translation by (c_x, c_y) (principal point)

Maps 3D rays to 2D pixels

 \mathcal{O} : aspect ratio (1 unless pixels are not square)

S: skew (0 unless pixels are shaped like rhombi/parallelograms)

 (c_x,c_y) : principal point ((w/2,h/2) unless optical axis doesn't intersect projection plane at image center)

Coordinate frames

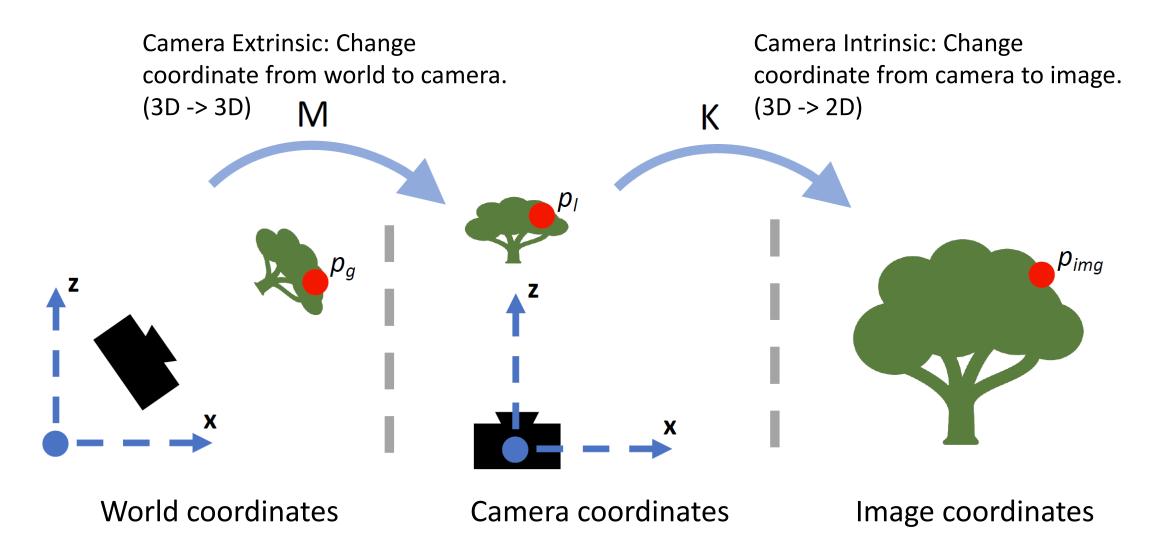


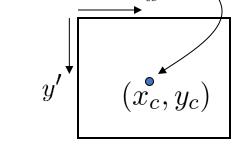
Figure credit: Peter Hedman

Camera parameters

A camera is described by several parameters

- Translation T of the optical center from the origin of world coords
- Rotation R of the image plane
- focal length f, principal point (c_x, c_y) , pixel aspect size α
- blue parameters are called "extrinsics," red are "intrinsics"

Projection equation



identity matrix

- The projection matrix models the cumulative effect of all parameters
- Useful to decompose into a series of operations

$$\boldsymbol{\Pi} = \begin{bmatrix} f & s & c_x \\ 0 & \alpha f & c_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{R}_{3\times3} & \mathbf{0}_{3\times1} \\ \mathbf{0}_{1\times3} & 0 \end{bmatrix} \begin{bmatrix} \mathbf{I}_{3\times3} & \mathbf{T}_{3\times1} \\ \mathbf{0}_{1\times3} & 0 \end{bmatrix}$$
intrinsics
projection
rotation
translation

- The definitions of these parameters are **not** completely standardized
 - especially intrinsics—varies from one book to another

General pinhole camera matrix

$$\mathbf{P} = \mathbf{K}[\mathbf{R}|\mathbf{t}]$$

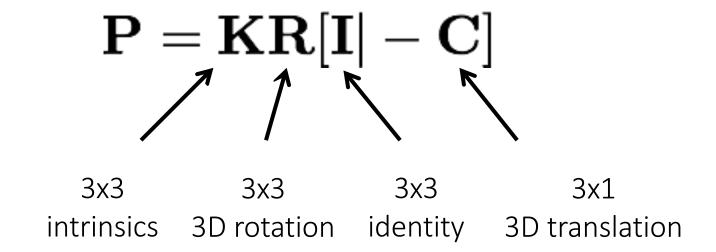
$$\mathbf{P} = \left[egin{array}{ccc|c} f & 0 & p_x \ 0 & f & p_y \ 0 & 0 & 1 \end{array}
ight] \left[egin{array}{ccc|c} r_1 & r_2 & r_3 & t_1 \ r_4 & r_5 & r_6 & t_2 \ r_7 & r_8 & r_9 & t_3 \end{array}
ight]$$
 intrinsic extrinsic parameters parameters

$$\mathbf{R} = \left[egin{array}{cccc} r_1 & r_2 & r_3 \ r_4 & r_5 & r_6 \ r_7 & r_8 & r_9 \end{array}
ight] \hspace{5mm} \mathbf{t} = \left[egin{array}{cccc} t_1 \ t_2 \ t_3 \end{array}
ight]$$

3D rotation 3D translation

Recap

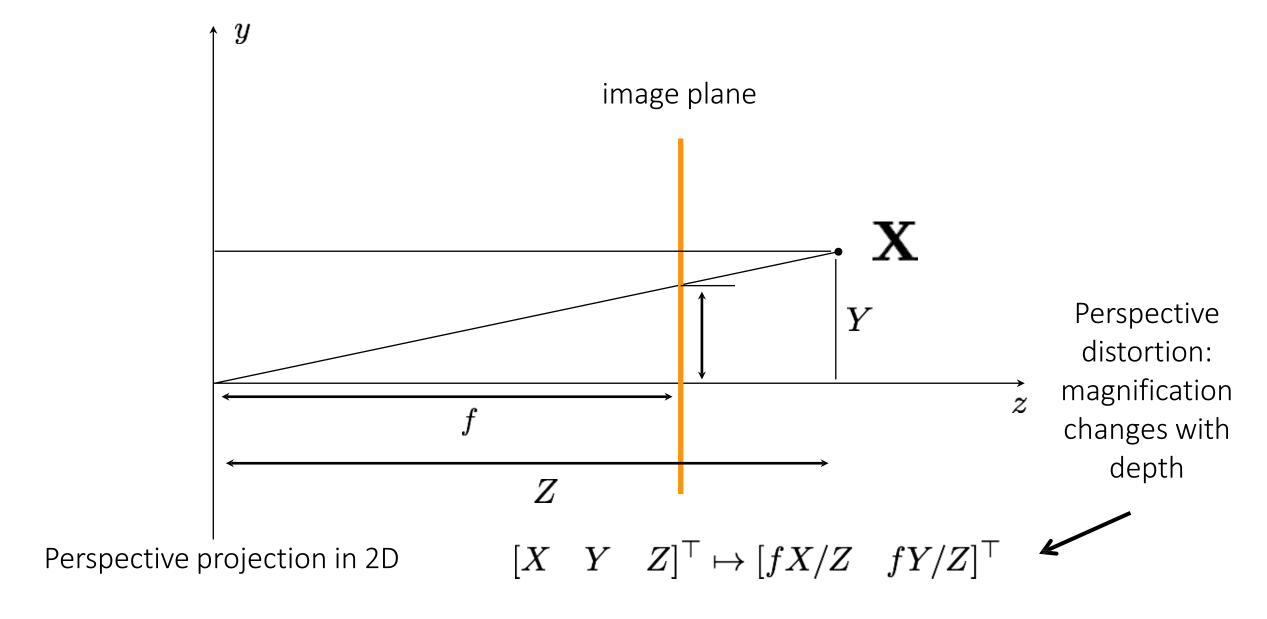
What is the size and meaning of each term in the camera matrix?



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- Other Projection models
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The 2D view of the (rearranged) pinhole camera



Perspective distortion



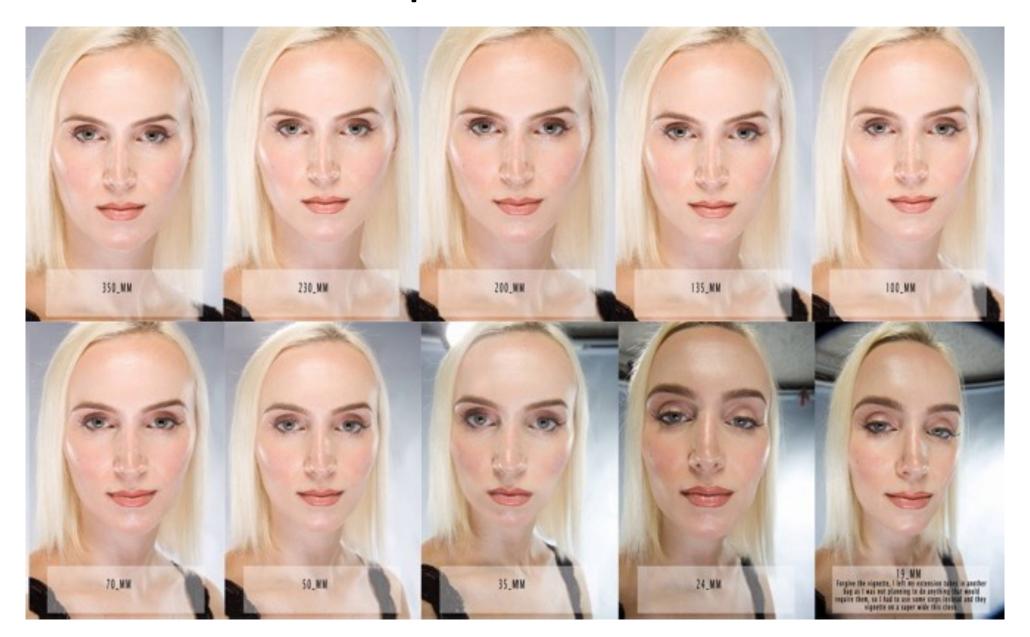


long focal length

mid focal length

short focal length

Perspective distortion





http://petapixel.com/2013/01/11/how-focal-length-affects-your-subjects-apparent-weight-as-seen-with-a-cat/

Forced perspective

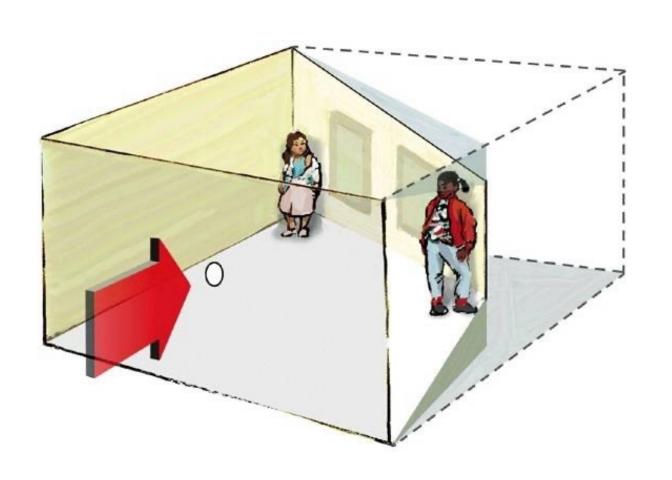


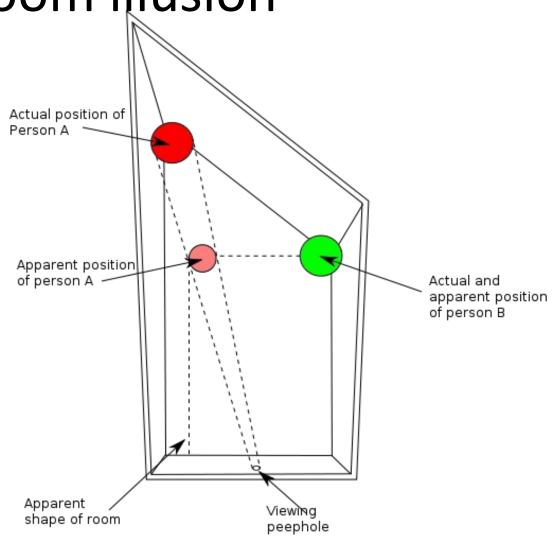


The Ames room illusion

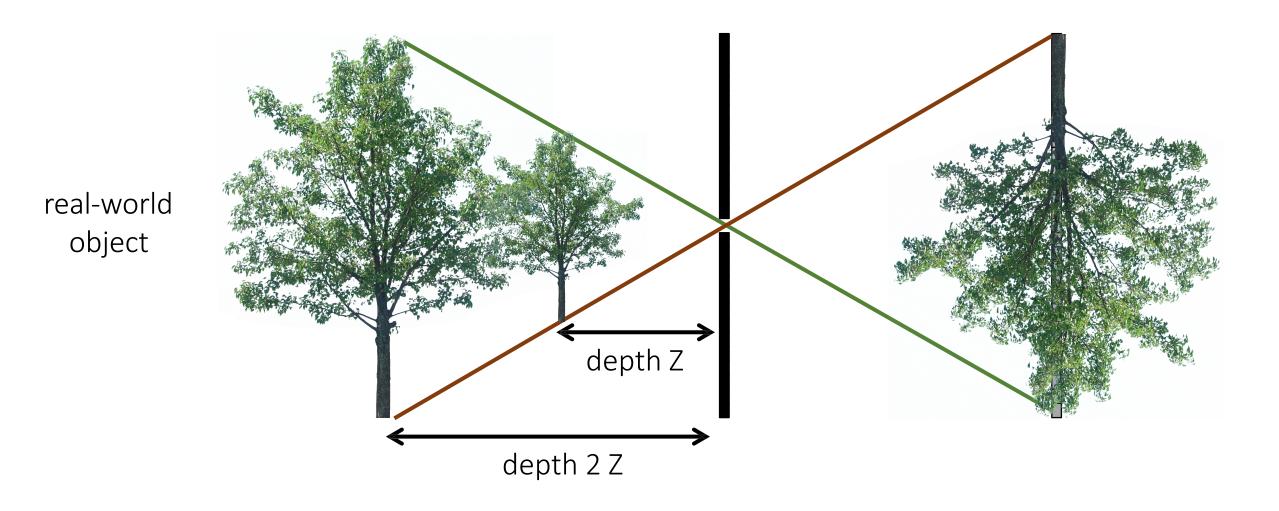


The Ames room illusion





Magnification depends on depth



Forced Perspective in movies (VFX)





Projective geometry—what's it good for?

- Uses of projective geometry
 - Drawing
 - Measurements
 - Mathematics for projection
 - Undistorting images
 - Camera pose estimation



Paolo Uccello

Dolly Zoom aka Vertigo Effect





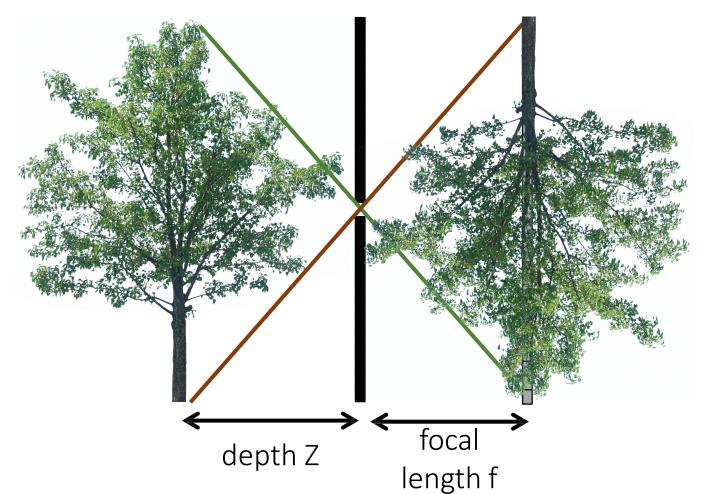
Other camera models

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What if...

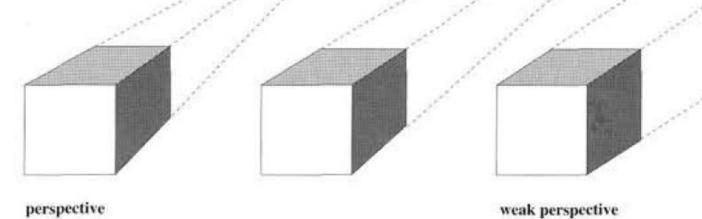
real-world object



... we continue increasing Z and f while maintaining same magnification?

$$f \to \infty$$
 and $\frac{f}{Z} = \text{constant}$

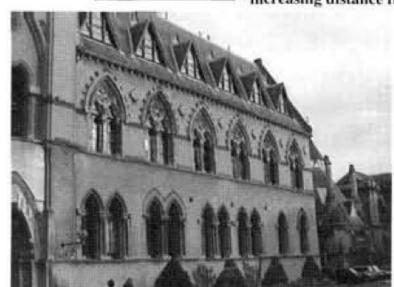
camera is *close* to object and has *small* focal length

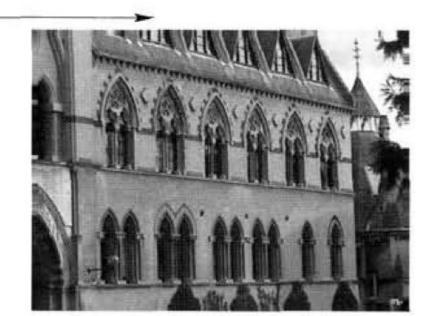


camera is *far* from object and has *large* focal length

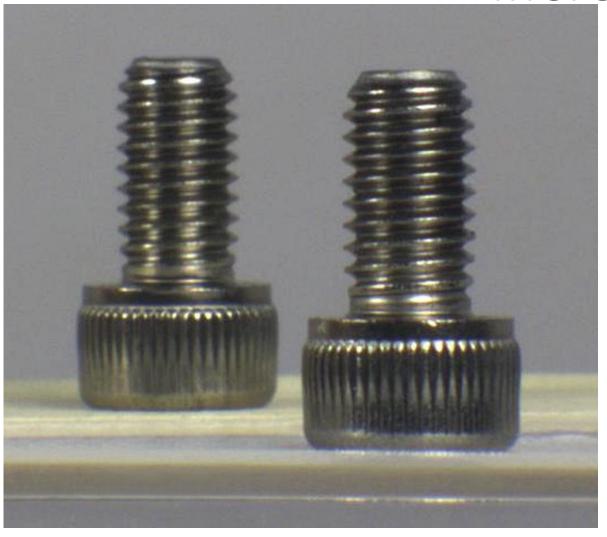
increasing focal length

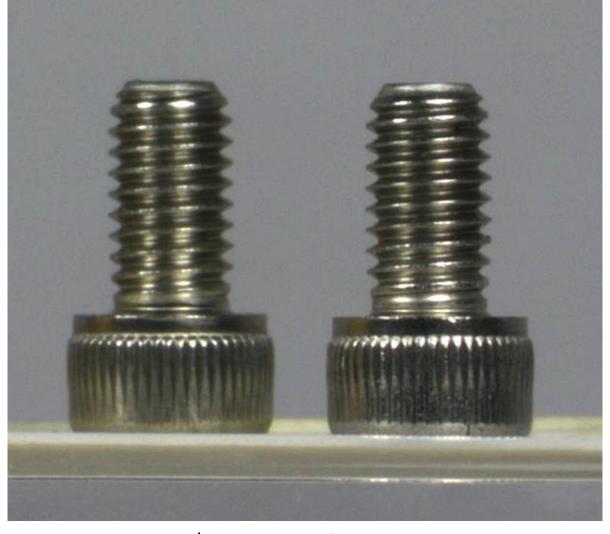
increasing distance from camera





Different cameras

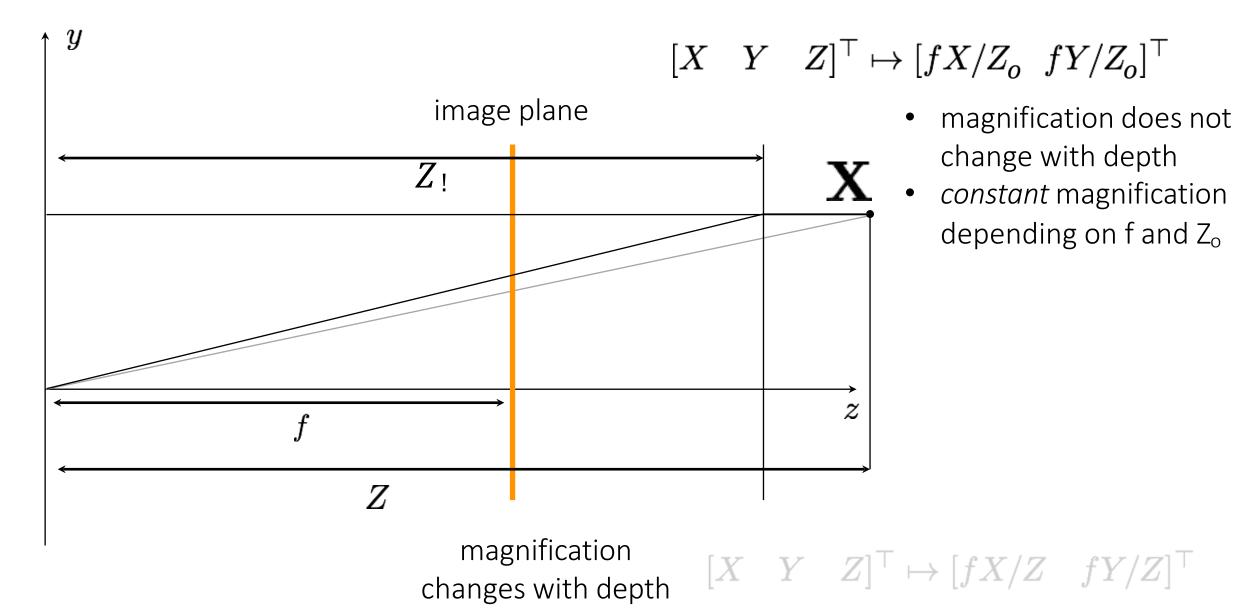




perspective camera

weak perspective camera

Weak perspective vs perspective camera



Comparing camera matrices

Let's assume that the world and camera coordinate systems are the same.

• The perspective camera matrix can be written as:

$$\mathbf{P} = \left[egin{array}{cccc} f & 0 & p_x \ 0 & f & p_y \ 0 & 0 & 1 \end{array}
ight] \left[egin{array}{cccc} 1 & 0 & 0 & 0 \ 0 & 1 & 0 & 0 \ 0 & 0 & 1 & 0 \end{array}
ight]$$

• The weak perspective camera matrix can be written as:

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1/d \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \begin{bmatrix} x \\ y \\ 1/d \end{bmatrix} \Rightarrow (dx, dy) \qquad \mathbf{P} = \begin{bmatrix} f & 0 & p_x \\ 0 & f & p_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & Z_o \end{bmatrix}$$

When can we assume a weak perspective camera?

When the scene (or parts of it) is very far away.



Weak perspective projection applies to the mountains.

When can we assume a weak perspective camera?

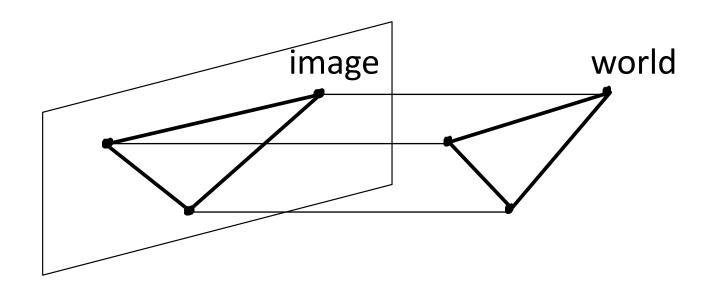
sensor distance D'

When we use a telecentric lens.

Place a pinhole at focal length, so that only rays parallel to primary ray pass through. focal length f object distance D

Orthographic camera

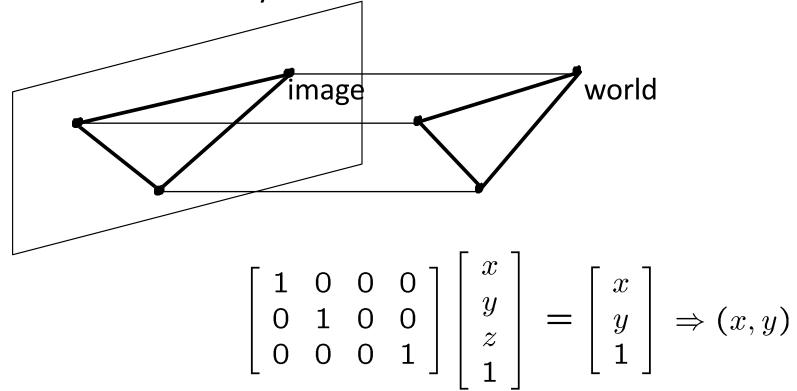
- Special case of weak perspective camera where:
- constant magnification is equal to 1.
- there is no shift between camera and image origins.
- the world and camera coordinate systems are the same.



What is the camera matrix in this case?

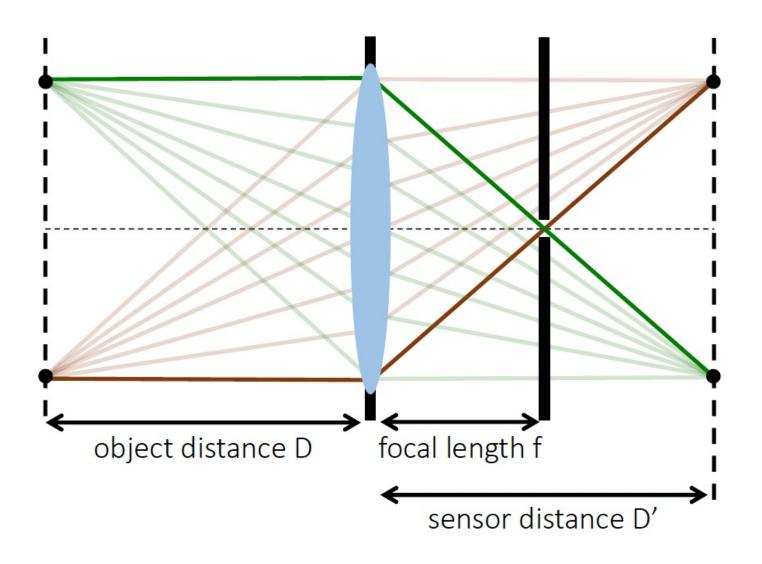
Orthographic camera

- Special case of weak perspective camera where:
- constant magnification is equal to 1.
- there is no shift between camera and image origins.
- the world and camera coordinate systems are the same.



Orthographic projection using a telecentric lens

How do we make the telecentric lens act as an orthographic camera?

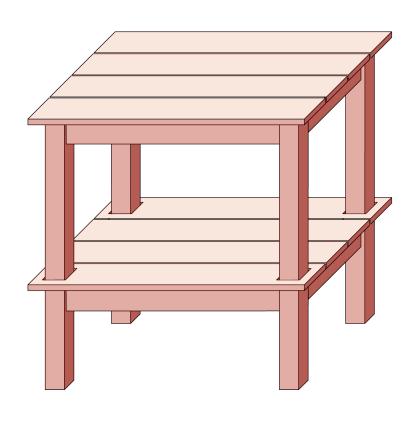


We set the sensor distance as:

$$D' = 2f$$

in order to achieve unit magnification.

Orthographic projection

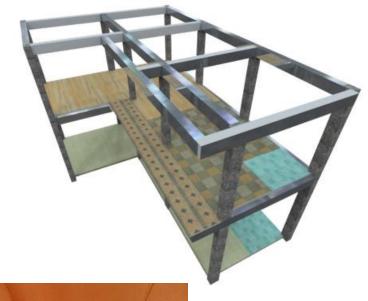






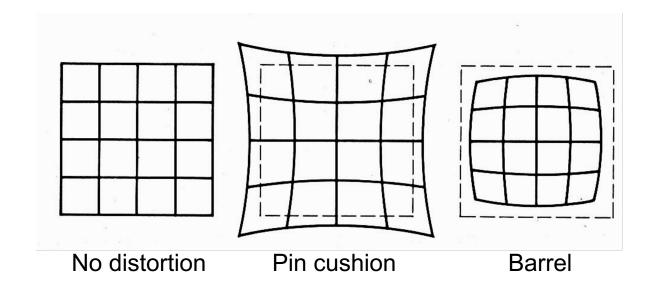
Perspective projection







Distortion

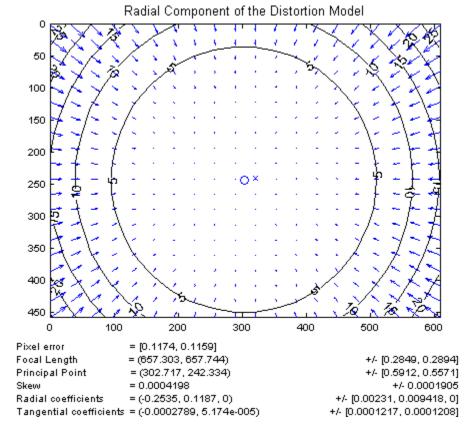


- Radial distortion of the image
 - Caused by imperfect lenses
 - Deviations are most noticeable for rays that pass through the edge of the lens





Radial distortion



 Arrows show motion of projected points relative to an ideal (distortion-free lens)

[Image credit: J. Bouguet http://www.vision.caltech.edu/bouguetj/calib_doc/htmls/example.html]

Correcting radial distortion





from <u>Helmut Dersch</u>

Today's Class

- Pinhole & Lens Camera
- Camera Parameters
 - Extrinsic
 - Intrinsic
- Perspective Distortion
- Other Projection models
- How to calibrate camera, i.e. estimate camera parameters (next class)

Slide Credits

- <u>CS5670, Introduction to Computer Vision</u>, Cornell Tech, by Noah Snavely.
- <u>CS 194-26/294-26</u>: Intro to Computer Vision and Computational Photography, UC Berkeley, by Angjoo Kanazawa.
- CS 16-385: Computer Vision, CMU, by Matthew O'Toole

Additional Reading

• Multiview Geometry, Hartley & Zisserman, Chapter 6.1, 6.2.

Related Readings from the past

Multiview Geometry, Hartley & Zisserman, Chapter 2.1, 2.2, 2.3, 2.4, 2.5, 2.6, 2.7, 3.1, 3.2, 4.1, 4.2