

Camera Models

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Some slides of this lecture are courtesy Silvio Savarese

A Camera in the 3D World



PyBullet with a Camera



Pinhole Camera



Pinhole Camera



Cannot be implemented in practice Useful for theoretic analysis

Central Projection in Camera Coordinates



Yu Xiang

Central Projection with Homogeneous Coordinates



Central projection

Principal Point Offset



Principle point: projection of the camera center



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From Metric to Pixels



Pixels, bottom-left coordinate systems

From Metric to Pixels

• Metric space, i.e., meters $\int f$

• Pixel space

$$\begin{bmatrix} f & p_y & 0 \\ & 1 & 0 \end{bmatrix}$$
$$\begin{bmatrix} \alpha_x & x_0 & 0 \\ & \alpha_y & y_0 & 0 \\ & & 1 & 0 \end{bmatrix}$$
$$\begin{bmatrix} \alpha_x = fm_x \\ & \alpha_y = fm_y \\ & x_0 = p_x m_x \end{bmatrix}$$

 $p_x \quad 0 \mid$

 m_x, m_y Number of pixel per unit distance

 $y_0 = p_y m_y$



The skew parameter will be zero for most normal cameras.

$$\begin{bmatrix} \alpha_x & x_0 & 0 \\ & \alpha_y & y_0 & 0 \\ & & 1 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} \longrightarrow \begin{bmatrix} \alpha_x \frac{x}{z} + x_0 \\ \alpha_y \frac{y}{z} + y_0 \end{bmatrix}$$

$$\begin{bmatrix} \alpha_x & -\alpha_x \cot(\theta) & x_0 & 0 \\ & \frac{\alpha_y}{\sin(\theta)} & y_0 & 0 \\ & & 1 & 0 \end{bmatrix}$$

https://blog.immenselyhappy.com/post/camera-axis-skew/

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Camera Intrinsics

$$\begin{bmatrix} \alpha_x & -\alpha_x \cot(\theta) & x_0 & 0 \\ & \frac{\alpha_y}{\sin(\theta)} & y_0 & 0 \\ & 1 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$
Camera intrinsics
$$K = \begin{bmatrix} \alpha_x & s & x_0 \\ & \alpha_y & y_0 \\ & 1 \end{bmatrix} \quad \begin{array}{c} \mathbf{x} = K \begin{bmatrix} I | \mathbf{0} \end{bmatrix} \mathbf{X}_{\text{cam}}$$

$$K = \begin{bmatrix} \alpha_y & y_0 \\ & 1 \end{bmatrix} \quad \begin{array}{c} \mathbf{x}_{11} & \mathbf{x}_{12} \end{bmatrix} \quad \begin{array}{c} \mathbf{x}_{12} = K \begin{bmatrix} I | \mathbf{0} \end{bmatrix} \mathbf{X}_{\text{cam}}$$

Homogeneous coordinates

Camera Extrinsics: Camera Rotation and Translation



Camera Projection Matrix $\,P = K[R|\mathbf{t}]\,$

• Homogeneous coordinates



Back-projection to a Ray in the World Coordinate



• The camera center O is on the ray

$$P^+\mathbf{x}\;$$
 is on the ray

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

Pseudo-inverse

The ray can be written as

 $\mathbf{X}(\lambda) = (1 - \lambda)P^{+}\mathbf{x} + \lambda O$

A pixel on the image backprojects to a ray in 3D

Back-projection to a 3D Point in Camera Coordinates



Back-projection to a 3D Point in Camera Coordinates





Equivalently

 $\mathbf{x} = K[I|\mathbf{0}]\mathbf{X}_{\text{cam}}$ $K^{-1}\mathbf{x}$ 3D point with depth $d: dK^{-1}\mathbf{X}$

The Pinhole Camera Model

• Camera projection matrix: intrinsics and extrinsics



Aperture Size of Pinhole Camera



What happen if the aperture is too small?

- Less light passes through
- Adding lenses



Lenses



Figure 4.8: (a) The earliest known artificially constructed lens, which was made between 750 and 710 BC in ancient Assyrian Nimrud. It is not known whether this artifact was purely ornamental or used to produce focused images. Picture from the British Museum. (b) A painting by Conrad con Soest from 1403, which shows the use of reading glasses for an elderly male.

Snell's Law

• How much rays of light bend when entering and exiting a transparent material

 θ_1

 n_1

 n_2

 S^{\bullet}

- Refractive index of a material $\,n=\,$
 - Air n = 1.000293, water n = 1.33
 - Crown glass n = 1.523

 $n_1 < n_2$

• Snell's Law

 $n_1 \sin \theta_1 = n_2 \sin \theta_2$

Speed of light in the medium

$$\theta_2 = \sin^{-1} \left(\frac{n_1 \sin \theta_1}{n_2} \right)$$

 θ_2

Convex Lenses

• Prisms

• A simple convex lens



Convex Lenses

- Objects in distance
 - Cameras



- Objects very close
 - Magnification
 - VR headsets



Controllable Aperture

- In the pinhole case, all depths are "in focus", but there may not enough lights
- When using a convex lens, it focuses objects at a single depth





Figure 4.34: A spectrum of aperture settings, which control the amount of light that enters the lens. The values shown are called the *focal ratio* or *f-stop*.

Shutters

• Collecting photons for each pixel





CMOS sensors



• Rolling shutter vs. Global shutter



Global shutter





Rolling shutter

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Chromatic Aberration

- The speed of light through a medium depends on the wavelength
 - Solution: combining convex and concave lenses of different materials



Figure 4.17: Chromatic aberration is caused by longer wavelengths travelin quickly through the lens. The unfortunate result is a different focal plane for wavelength or color.



Figure 4.18: The upper image is properly focused whereas the lower image suffers from chromatic aberration. (Figure by Stan Zurek, license CC-BY-SA-2.5.)

Spherical Aberration

• Rays further away from the lens center being refracted more than rays near the center





Aspheric lens

Optical Distortion



 The variation of refractive index towards the outer extremities of a rotational symmetric lens can cause magnification changes in the image space, depending on the distance from the principal axis.



Angular Field of View (AFOV)

$$\mathrm{AFOV} = 2 imes an^{-1} \left(rac{H}{2f}
ight)$$



Figure 1: For a given sensor size, H, shorter focal lengths produce wider AFOV's.

Barrel Distortion of Fisheye Cameras



Figure 4.21: An image with barrel distortion, taken by a fish-eyed lens. (Image by Wikipedia user Ilveon.)

Tangential Distortion

• Camera sensor mis-alignment during the manufacturing process



Distortion Correction

• The Brown-Conrady distortion model [Wikipedia]

$$egin{aligned} x_{\mathrm{u}} &= x_{\mathrm{d}} + (x_{\mathrm{d}} - x_{\mathrm{c}})(K_{1}r^{2} + K_{2}r^{4} + \cdots) + (P_{1}(r^{2} + 2(x_{\mathrm{d}} - x_{\mathrm{c}})^{2}) \ &+ 2P_{2}(x_{\mathrm{d}} - x_{\mathrm{c}})(y_{\mathrm{d}} - y_{\mathrm{c}}))(1 + P_{3}r^{2} + P_{4}r^{4} \cdots) \ &y_{\mathrm{u}} &= y_{\mathrm{d}} + (y_{\mathrm{d}} - y_{\mathrm{c}})(K_{1}r^{2} + K_{2}r^{4} + \cdots) + (2P_{1}(x_{\mathrm{d}} - x_{\mathrm{c}})(y_{\mathrm{d}} - y_{\mathrm{c}}) \ &+ P_{2}(r^{2} + 2(y_{\mathrm{d}} - y_{\mathrm{c}})^{2}))(1 + P_{3}r^{2} + P_{4}r^{4} \cdots), \end{aligned}$$



where

- $(x_{
 m d}, \, y_{
 m d})$ is the distorted image point as projected on image plane using specified lens;
- $(x_{\mathrm{u}}, \, y_{\mathrm{u}})$ is the undistorted image point as projected by an ideal pinhole camera;
- $(x_{
 m c},\ y_{
 m c})$ is the distortion center;
- K_n is the $n^{ ext{th}}$ radial distortion coefficient;

Use calibration tools to estimate these coefficients

• P_n is the $n^{ ext{th}}$ tangential distortion coefficient; and

• $r = \sqrt{(x_{\rm d} - x_{\rm c})^2 + (y_{\rm d} - y_{\rm c})^2}$, the Euclidean distance between the distorted image point and the distortion center.^[3]

Summary: Camera Models

• Camera projection matrix: intrinsics and extrinsics

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

- Lens distortion
 - Radial distortion coefficients K_1, K_2, K_3, \ldots
 - Tangential distortion coefficients P_1, P_2, P_3, \ldots

Further Reading

- Section 2.1, Computer Vision, Richard Szeliski
- Multiview Geometry in Computer Vision, Richard Hartley and Andrew Zisserman, Chapter 6, Camera Models
- Stanford CS231A: Computer Vision, From 3D Reconstruction to Recognition, Lecture 2 <u>https://web.stanford.edu/class/cs231a/syllabus.html</u>
- Image formation by lenses https://courses.lumenlearning.com/physics/chapter/25-6-image-formation-bylenses/
- Distortion (Wikipedia) https://en.wikipedia.org/wiki/Distortion (optics)