

Institute of Informatics - Institute of Neuroinformatics



#### Lecture 02 Image Formation 1

Davide Scaramuzza http://rpg.ifi.uzh.ch

#### Lab Exercise 1 - Today afternoon

- Room ETH HG E 1.1 from 13:15 to 15:00
- > Work description: implement an augmented reality wireframe cube
  - Practice the perspective projection



### **Outline of this lecture**

- Image Formation
- Other camera parameters
- Digital camera
- Perspective camera model
- Lens distortion

#### Historical context

- Pinhole model: Mozi (470-390 BCE), Aristotle (384-322 BCE)
- Principles of optics (including lenses): Alhacen (965-1039)
- Camera obscura: Leonardo da Vinci (1452-1519), Johann Zahn (1631-1707)
- First photo: Joseph Nicephore Niepce (1822)
- Daguerréotypes (1839)
- Photographic film (Eastman, 1888, founder of Kodak)
- Cinema (Lumière Brothers, 1895)
- Color Photography (Lumière Brothers, 1908)
- Television (Baird, Farnsworth, Zworykin, 1920s)
- First consumer camera with CCD: Sony Mavica (1981)
- First fully digital camera: Kodak DCS100 (1990)







Niepce, "La Table Servie," 1822



CCD chip

## Image formation

How are objects in the world captured in an image?



#### How to form an image



Place a piece of film in front of an object
 ⇒ Do we get a reasonable image?

#### Pinhole camera



- Add a barrier to block off most of the rays
  - This reduces blurring
  - The opening is known as the **aperture**

#### Camera obscura

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Sic nos exacté Anno.1544. Louanii ecliphim Solis obferuauimus, inuenimusq; deficere paulò plus g dexIn Latin, means 'dark room'

- Basic principle known to Mozi (470-390 BC), Aristotle (384-322 BC)
- Drawing aid for artists: described by Leonardo da Vinci (1452-1519)
- Image is inverted
- Depth of the room (box) is the effective focal length

"Reinerus Gemma-Frisius, observed an eclipse of the sun at Louvain on January 24, 1544, and later he used this illustration of the event in his book <u>De Radio Astronomica et</u> <u>Geometrica</u>, 1545. It is thought to be the first published illustration of a camera obscura..." Hammond, John H., <u>The Camera Obscura, A Chronicle</u>

#### Camera obscura at home



Figure 1 - A lens on the window creates the image of the external world on the opposite wall and you can see it every morning, when you wake up.



#### Home-made pinhole camera



What can we do to reduce the blur?

#### Effects of the Aperture Size

- In an ideal pinhole, only one ray of light reaches each point on the film ⇒ the image can be very dim
- Making the aperture bigger makes the image blurry



#### Shrinking the aperture



Why not make the aperture as small as possible?

#### Shrinking the aperture



Why not make the aperture as small as possible?

- Less light gets through (must increase the exposure)
- Diffraction effects...

#### Image formation using a converging lens



- A lens focuses light onto the film
- Rays passing through the Optical Center are not deviated

#### Image formation using a converging lens



 All rays parallel to the Optical Axis converge at the Focal Point



#### Find a relationship between *f*, *z*, and *e*

#### Thin lens equation



- Similar Triangles:  $\frac{B}{A} = \frac{e}{z}$   $\frac{B}{A} = \frac{e-f}{f} = \frac{e}{f} - 1$   $\frac{B}{A} = \frac{e-f}{f} = \frac{e}{f} - 1$ • Any object point satisfying this equation is in focus
  - Can I use this to measure distances?

"In focus"



- For a fixed film distance from the lens, there is a specific distance between the object and the lens, at which the object appears "in focus" in the image
- Other points project to a "blur circle" in the image

#### **Blur Circle**



- Object is out of focus  $\Rightarrow$  Blur Circle has radius:  $R = \frac{L\delta}{2e}$ 
  - A small *L* (pinhole) gives a small *R* (Blur Circle)
  - To capture a 'good' image: adjust camera settings, such that R remains smaller than the image resolution

#### The Pin-hole approximation



• We need to adjust the image plane such that objects at infinity are in focus. As the object gets far, the image plane gets closer to the focal plane

$$\frac{1}{f} = \frac{1}{z} - \frac{1}{e} \implies \frac{1}{f} \approx \frac{1}{e} \implies f \approx e$$
$$\cong 0$$

#### The Pin-hole approximation



What happens if  $z \gg f$  and  $z \gg L$ ?

We need to adjust the image plane such that objects at infinity are in focus. As the object gets far, the image plane gets closer to the focal plane

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$$\cong 0$$

#### The Pin-hole approximation

• What happens if  $z \gg f$  and  $z \gg L$ ?



- This is known as Pinhole Approximation and the relation between the image and object becomes:  $\frac{h'}{h} = \frac{f}{h} \Rightarrow h' = \frac{f}{h}h$
- h = z = zThe dependence of the image of an object on its depth (i.e. distance from the camera) is known as **perspective**

#### Perspective effects

• Far away objects appear smaller



#### Perspective effects



#### Perspective and art

- Use of correct perspective projection observed in 1<sup>st</sup> century BC frescoes
- During Renaissance time, artists developed systematic methods to determine perspective projection (around 1480-1515)





Raphael

# **Playing with Perspective**

- Perspective gives us very strong depth cues
   ⇒ hence we can perceive a 3D scene by viewing its 2D representation (i.e. image)
- An example where perception of 3D scenes is misleading is the Ames room (check out the Ames room in the Technorama science museum in Winterthur)



#### "Ames room"

A clip from "The computer that ate Hollywood" documentary. Dr. Vilayanur S. Ramachandran. https://youtu.be/Ttd0YjXF0no

#### **Perspective Projection**

What is preserved?

• Straight lines are still straight



#### **Perspective Projection**

What is lost?

- Length
- Angles



#### Vanishing points and lines

Parallel lines in the world intersect in the image at a "vanishing point"



## Vanishing points and lines

Parallel lines in the world intersect in the image at a "vanishing point" Parallel planes in the world intersect in the image at a "vanishing line"



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#### Focus and Depth of Field

- Depth of Field (DOF) is the distance between the nearest and farthest objects in a scene that appear acceptably sharp in an image.
- Although a lens can precisely focus at only one distance at a time, the decrease in sharpness is gradual on each side of the focused distance, so that within the DOF, the unsharpness is imperceptible under normal viewing conditions



*are using.* If you the the depth of field will be to infinity. <a href="https://www.sec.example.com">For infinity.</a> For infinity. <a href="https://www.sec.example.com">For infinity.</a>

## Focus and Depth of Field

• How does the aperture affect the depth of field?



• A smaller aperture increases the DOF but reduces the amount of light into the camera

# Field of View (FOV)

Angular measure of portion of 3D space seen by the camera







70 mm lens, 28.9° × 19.5°



210 mm lens, 9.8° × 6.5°

#### Field of view depends on focal length

- As *f* gets smaller, image becomes more wide angle
  - more world points project onto the finite image plane
- As *f* gets larger, image becomes more *narrow* angle
  - smaller part of the world projects onto the finite image plane



#### Relation between field of view and focal length



$$\tan \frac{\theta}{2} = \frac{W}{2f} \quad \text{or} \quad f = \frac{W}{2} \left[ \tan \frac{\theta}{2} \right]^{-1}$$

Smaller FOV = larger Focal Length

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### **Digital cameras**

• The film is a array of CCD or CMOS light sensitive diodes that convert photons (light energy) into electrons



# Digital images

Pixel Intensity with 8 bits ranges between [0,255]



j=1

NB. Matlab coordinates: [rows, cols]; C/C++ [cols, rows]

width

500

i=1

## Color sensing in digital cameras

Bayer grid



- The Bayer pattern (invented by Bayer in 1976, who worked at Kodak) places green filters over half of the sensors (in a checkerboard pattern), and red and blue filters over the remaining ones.
- This is because the luminance signal is mostly determined by green
   values and the human visual system is much more sensitive to high frequency detail in luminance than in chrominance.

## Color sensing in digital cameras

Bayer grid



For each pixel, estimate missing color components from neighboring values (demosaicing)



Foveon chip design (http://www.foveon.com) stacks the red, green, and blue sensors beneath each other but has not gained widespread adoption.

#### Color sensing in digital cameras

#### RGB color space

... but there are also many other color spaces... (e.g., YUV)









#### **Rolling vs Global Shutter Camera**





#### **Rolling Shutter**

- Rows of pixels are exposed at different times, one after the other
- No problem for still or slow objects
- But may distort image for moving objects

#### **Global Shutter**

- All pixels are exposed simultaneously
- Good for moving objects
- No image distortion

### Rolling vs Global Shutter Camera

- Rolling Shutter cameras may distort (skew) moving objects
- Global Shutter cameras do not have this problem



**Rolling shutter** 

**Global shutter** 

#### Cool applications of rolling shutter cameras

High speed wave-form capturing from a guitar



Recorded from an iPhone 4 (rolling shutter): <u>https://youtu.be/TKF6nFzpHBU</u>

#### An example camera datasheet

#### **MVBlueFOX-IGC / -MLC**

**Technical Details** 



mvBlueFO mvBlueFO	X-IGC X-MLC	Resolution (H x V pixels)	Sensor size (optical)	Pixel size (µm)	Frame rate	Sensor technology	Readout type	ADC resolution / output in bits	Sensor
-200w <sup>12</sup>	G/C	752 x 480	1/3″	6 x 6	90	CMOS	Global	$10 \rightarrow 10/8$	Aptina MT9V
-202b	G/C	1280 x 960	1/3″	3.75 x 3.75	24.6	CMOS	Global	10→10/8	Aptina MT9M
-202d <sup>1</sup>	G/C	1280 x 960	1/3"	3.75 x 3.75	24.6	CMOS	Rolling	10→10/8	Aptina MT9M
-205 <sup>2</sup>	G/C	2592 x 1944	1/2.5"	2.2 x 2.2	5.8	CMOS	Global Reset	$10 \rightarrow 10/8$	Aptina MT9P

#### High Dynamic Range (HDR) mode supported

Software trigger supported

Sample: mvBlueFOX-IGC200wG means version with housing and 752 x 480 CMOS gray scale sensor. mvBlueFOX-MLC200wG means single-board version without housing and with 752 x 480 CMOS gray scale sensor.

#### Hardware Features

Gray scale / Color	Gray scale (G) / Color (C)	
Interface	USB 2.0 (up to 480 Mbit/s)	
Image formats	Mono8, Mono10, BayerGR8, BayerGR10	

Triggers	Externa	External hardware based (optional), software based (depending on the sensor) or free run				
Size w/o lens (W x H	x L)   Weight w/o lens	mvBlueFOX-IGC: mvBlueFOX-MLC:	39.8 x 39.8 x 16.5 mm   approx. 10 g 35 x 33 x 25 mm (without lens mount)   approx. 80 g	i		
Permissible ambient temperature		Operation: Storage:	045 °C / 30 to 80 % RH -2060 °C / 20 to 90 % RH			
Lens mounts	Back fo	focus adjustable C/CS-mount lens holder / C-mount, CS-mount or optional S-mount				
Digital I/Os	mvBlueFOX-IGC (optional) mvBlueFOX-MLC	1 / 1 opt 1 / 1 opt	o-isolated o-isolated or 2 / 2 TTL compliant			
Conformity	CE, FCC, RoHS					
Driver	mvIMPACT Acquire SDK					
Operating systems		Windows®, Linux® - 32 bit and 64 bit				



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#### **Perspective Camera**



- For convenience, the image plane is usually represented in front of C such that the image preserves the same orientation (i.e. not flipped)
- Note: a camera does not measure distances but angles!
   a camera is a "bearing sensor"

#### From World to Pixel coordinates

Goal: Find pixel coordinates (u, v) of point  $P_w$  in the world frame:

- 1. Convert world point  $P_w$  to camera point  $P_c$  through rigid body transform [R, T]
- 2. Convert  $P_c$  to image-plane coordinates (x,y)
- 3. Convert (*x*,*y*) to (discretized) pixel coordinates (*u*,*v*)



#### Perspective Projection (1)

#### From the Camera frame to the image plane



- The Camera point  $P_c = (X_c, 0, Z_c)^T$  projects to p = (x, y) onto the image plane
- From similar triangles:  $\frac{x}{f} = \frac{X_c}{Z_c} \Rightarrow x = \frac{fX_c}{Z_c}$
- Similarly, in the general case:



#### Perspective Projection (2)

#### From the Camera frame to pixel coordinates

- To convert p from the local image plane coords (x,y) to the pixel coords (u,v), we need to account for:
  - the pixel coords of the camera optical center  $O = (u_0, v_0)$
  - Scale factors  $k_u$ ,  $k_v$  for the pixel-size in both dimensions

So:  

$$u = u_0 + k_u x \Longrightarrow u = u_0 + \frac{k_u f X_c}{Z_c}$$

$$v = v_0 + k_v y \Longrightarrow v = v_0 + \frac{k_v f Y_c}{Z_c}$$



Use Homogeneous Coordinates for linear mapping from 3D to 2D, by introducing an extra element (scale):

$$p = \begin{pmatrix} u \\ v \end{pmatrix} \qquad \qquad \widetilde{p} = \begin{bmatrix} \widetilde{u} \\ \widetilde{v} \\ \widetilde{w} \end{bmatrix} = \lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix}$$

#### Perspective Projection (3)

So:

$$v = v_0 + \frac{k_v f Y_c}{Z_c}$$

 $u = u_c + \frac{k_u f X_c}{k_c}$ 

Expressed in matrix form and homogeneous coordinates:

$$\begin{bmatrix} \lambda u \\ \lambda v \\ \lambda \end{bmatrix} = \begin{bmatrix} k_u f & 0 & u_0 \\ 0 & k_v f & v_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} X_c \\ Y_c \\ Z_c \end{bmatrix}$$

Or alternatively





K is called "Calibration matrix" or "Matrix of Intrinsic Parameters"

In the past it was common to assume a skew factor ( $K_{12} \neq 0$ ) to account for possible skew in the pixel manufacturing process. However, the camera manufacturing process today is so good that we can safely assume  $K_{12} = 0$  and  $\alpha_u = \alpha_v$ .

- Determine the Intrinsic Parameter Matrix (K) of a digital camera with image size  $640 \times 480$  pixels and a horizontal field of view of  $90^{\circ}$
- Assume that the principal point is the center of the diagonals
- Also, assume that we have square pixels
- What is the vertical field of view?

- Determine the Intrinsic Parameter Matrix (K) of a digital camera with image size  $640 \times 480$  pixels and a horizontal field of view of  $90^{\circ}$
- Assume that the principal point is the center of the diagonals
- Also, assume that we have square pixels
- What is the vertical field of view?



• Prove that world's parallel lines intersect at a vanishing point in the camera image



- Prove that world's parallel lines intersect at a vanishing point in the camera image
- Let's consider the perspective projection equation in camera metric coordinates:

$$x = f \frac{X}{Z}, \quad y = f \frac{Y}{Z}$$

• Two parallel 3D lines have parametric equations:

$$\begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = \begin{bmatrix} X_0 \\ Y_0 \\ Z_0 \end{bmatrix} + s \begin{bmatrix} l \\ m \\ n \end{bmatrix} \qquad \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = \begin{bmatrix} X_1 \\ Y_1 \\ Z_1 \end{bmatrix} + s \begin{bmatrix} l \\ m \\ n \end{bmatrix}$$

• Now substitute this into the camera perspective projection equation and compute the limit for  $s \to \infty$ 

$$\lim_{s \to \infty} f \frac{X_i + sl}{Z_i + sn} = f \frac{l}{n} = x_{VP}, \qquad \lim_{s \to \infty} f \frac{Y_i + sm}{Z_i + sn} = f \frac{m}{n} = y_{VP}$$

- The result solely depends on the direction vector of the line. These are the image coordinates of the vanishing point (VP).
- What is the intuitive interpretation of this?

#### Perspective Projection (4)

#### From the World frame to the Camera frame

$$\begin{bmatrix} X_{c} \\ Y_{c} \\ Z_{c} \end{bmatrix} = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} \begin{bmatrix} X_{w} \\ Y_{w} \\ Z_{w} \end{bmatrix} + \begin{bmatrix} t_{1} \\ t_{2} \\ t_{3} \end{bmatrix}$$

$$\begin{bmatrix} X_{c} \\ Y_{c} \\ Z_{c} \end{bmatrix} = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} \begin{bmatrix} t_{1} \\ t_{2} \\ t_{3} \end{bmatrix} = \begin{bmatrix} R \\ T \end{bmatrix} \cdot \begin{bmatrix} X_{w} \\ Y_{w} \\ Z_{w} \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} X_{w} \\ Y_{w} \\ Z_{w} \\ 1 \end{bmatrix} = \begin{bmatrix} R \\ T \end{bmatrix} \cdot \begin{bmatrix} X_{w} \\ Y_{w} \\ Z_{w} \\ 1 \end{bmatrix}$$

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#### Normalized image coordinates

In both computer vision and robotics, it is often convenient to use normalized image coordinates

- Let (u, v) be the pixel coordinates of an image point
- We define the normalized image coordinates  $(\bar{x}, \bar{y})$

$$\begin{bmatrix} \overline{u} \\ \overline{v} \\ 1 \end{bmatrix} = K^{-1} \begin{bmatrix} u \\ v \\ 1 \end{bmatrix}$$

• Normalized image coordinates can be interpreted as image coordinates on an virtual image plane with focal length equal to 1 unit (adimensional)

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#### **Radial Distortion**



No distortion

Positive radial distortion (Barrel distortion)  $k_1 > 0$  Negative radial distortion (Pincushion distortion)  $k_1 < 0$ 

#### **Radial Distortion**

- The standard model of radial distortion is a transformation **from the ideal coordinates** (*u*, *v*) (i.e., **non-distorted**) **to the real, observed coordinates** (distorted) (*u*<sub>d</sub>, *v*<sub>d</sub>)
- For a given non distorted image point (u, v), the amount of distortion is a nonlinear function of it distance r from principal point. For most lenses, a simple quadratic model of distortion produces good results

$$\begin{bmatrix} u_d \\ v_d \end{bmatrix} = (1+k_1r^2) \begin{bmatrix} u-u_0 \\ v-v_0 \end{bmatrix} + \begin{bmatrix} u_0 \\ v_0 \end{bmatrix}$$
$$r^2 = (u-u_0)^2 + (v-v_0)^2$$

where

Radial & Tangential Distortion in the OpenCV and Matlab Camera Models

- **Radial Distortion:** Depending on the amount of distortion (an thus on the camera field of view), higher order terms can be introduced for the radial distortion
- **Tangential Distortion**: if the lens is misaligned (not perfectly orthogonal to the image sensor), a *non-radial* distortion is introduced



The **left figure** shows the impact of the **complete distortion model (radial + tangential)** on each pixel of the image. Each arrow represents the effective displacement of a pixel induced by the lens distortion. Observe that points at the corners of the image are displaced by as much as **25 pixels**. The **center figure** shows the impact of the **tangential component of distortion**. On this plot, the **maximum induced displacement is 0.14 pixel** (at the upper left corner of the image). Finally, the **right figure** shows the impact of the **radial component of distortion**. This plot is very similar to the full distortion plot, showing that the tangential component could very well be discarded in the complete distortion model. On the three figures, the cross indicates the center of the image (i.e., the center of the two diagonals), and the circle the location of the principal point.

#### Summary: Perspective projection equations

• To recap, a 3D world point  $P = (X_w, Y_w, Z_w)$  projects into the image point p = (u, v)

$$\widetilde{p} = \begin{bmatrix} \widetilde{u} \\ \widetilde{v} \\ \widetilde{w} \end{bmatrix} = \lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = K \begin{bmatrix} R \mid T \end{bmatrix} \cdot \begin{bmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{bmatrix} \quad \text{where} \quad K = \begin{bmatrix} \alpha & 0 & u_0 \\ 0 & \alpha & v_0 \\ 0 & 0 & 1 \end{bmatrix}$$

and  $\lambda$  is the depth ( $\lambda = Z_C$ ) of the scene point

• If we want to take into account the radial distortion, then the distorted coordinates  $(u_d, v_d)$  (in pixels) can be obtained as

$$\begin{bmatrix} u_d \\ v_d \end{bmatrix} = (1 + k_1 r^2) \begin{bmatrix} u - u_0 \\ v - v_0 \end{bmatrix} + \begin{bmatrix} u_0 \\ v_0 \end{bmatrix}$$

where

$$r^2 = (u - u_0)^2 + (v - v_0)^2$$

• See also the OpenCV documentation:

http://docs.opencv.org/2.4.13.3/modules/calib3d/doc/camera\_calibration\_and\_3d\_reconstruction.html

## Summary (things to remember)

- Perspective Projection Equation
- Intrinsic and extrinsic parameters (K, R, t)
- Homogeneous coordinates
- Normalized image coordinates
- Image formation equations (including simple radial distortion)
- Chapter 4 of Autonomous Mobile Robot book: <u>http://rpg.ifi.uzh.ch/docs/teaching/2018/Ch4\_AMRobots.pdf</u>

## **Understanding Check**

Are you able to:

- Explain what a blur bircle is?
- Derive the thin lens equation and perform the pinhole approximation?
- Define vanishing points and lines?
- Prove that parallel lines intersect at vanishing points?
- Explain how to build an Ames room?
- Derive a relation between the field of view and the focal length?
- Explain the perspective projection equation, including lens distortion and world to camera projection?