## VC 13/14-T2 Image Formation

Mestrado em Ciência de Computadores
Mestrado Integrado em Engenharia de Redes e Sistemas Informáticos

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## Outline

## - ‘Computer Vision’?

- The Human Visual System
- Image Capturing Systems

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## Topic: Computer Vision?

- ‘Computer Vision’?
- The Human Visual System
- Image Capturing Systems


## Computer Vision

"The goal of Computer Vision is to make useful decisions about real physical objects and scenes based on sensed images",

Shapiro and Stockman, "Computer Vision", 2001


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## Components of a Computer Vision System



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## Topic: The Human Visual System

- ‘Computer Vision’?
- The Human Visual System
- Image Capturing Systems


## Our Eyes


-Iris is the diaphragm that changes the aperture (pupil)
-Retina is the sensor where the fovea has the highest resolution

## Focusing


shorter focal length
Changes the focal length of the lens

## Myopia and Hyperopia



## Astigmatism



The cornea is distorted causing images to be un-focused on the retina.

## Blind Spot in the Eye

Close your right eye and look directly at the " + "

## Colour

- Our retina has:
- Cones - Measure the frequency of light (colour)
- 6 to 7 millions
- High-definition
- Need high luminosity
- Rods - Measure the intensity of light (luminance)
- 75 to 150 millions
- Low-definition
- Function with low luminosity



## Topic: Image Capturing Systems

- ‘Computer Vision’?
- The Human Visual System
- Image Capturing Systems


## A Brief History of Images



1544

Camera Obscura, Gemma Frisius, 1544 U.PORTO

## A Brief History of Images



## A Brief History of Images



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## A Brief History of Images



## A Brief History of Images



## Components of a Computer Vision System



Scene Interpretation

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## Pinhole and the Perspective Projection



## Pinhole Camera

- Basically a pinhole camera is a box, with a tiny hole at one end and film or photographic paper at the other.
- Mathematically: out of all the light rays in the world, choose the set of light rays passing through a point and projecting onto a plane.



## Pinhole Photography


©Charlotte Murray Untitled, 4" x 5" pinhole photograph, 1992


Image Size inversely proportional to Distance
Reading: http://www.pinholeresource.com/

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## Magnification



From perspective projection:

$$
\begin{aligned}
& \frac{x^{\prime}}{f^{\prime}}=\frac{x}{z} \quad \frac{y^{\prime}}{f^{\prime}}=\frac{y}{z} \\
& \frac{x^{\prime}+\delta x^{\prime}}{f^{\prime}}=\frac{x+\delta x}{z} \quad \frac{y^{\prime}+\delta y^{\prime}}{f^{\prime}}=\frac{y+\delta y}{z}
\end{aligned}
$$

Magnification:

$$
\begin{gathered}
m=\frac{d^{\prime}}{d}=\frac{\sqrt{\left(\delta x^{\prime}\right)^{2}+\left(\delta y^{\prime}\right)^{2}}}{\sqrt{(\delta x)^{2}+(\delta y)^{2}}}=\frac{f^{\prime}}{z} \\
\frac{\text { Area }_{\text {image }}}{\text { Area }_{\text {scene }}}=m^{2}
\end{gathered}
$$

## Image Formation using Lenses

- Lenses are used to avoid problems with pinholes.
- Ideal Lens: Same projection as pinhole but gathers more light!

- Gaussian Thin Lens Formula: $\frac{1}{i}+\frac{1}{o}=\frac{1}{f}$
- $f$ is the focal length of the lens - determines the lens's ability to refract light


## Focus and Defocus



- Gaussian Law:

$$
\frac{1}{i}+\frac{1}{o}=\frac{1}{f} \quad \square \quad\left(i^{\prime}-i\right)=\frac{f}{\left(o^{\prime}-f\right)} \frac{f}{(o-f)}\left(o-o^{\prime}\right)
$$

$$
\frac{1}{i^{\prime}}+\frac{1}{o^{\prime}}=\frac{1}{f}
$$

- In theory, only one scene plane is in focus.


## Depth of Field

- Range of object distances over which image is sufficiently well focused.
- Range for which blur circle is less than the resolution of the sensor.

$\square$


## Chromatic Aberration


longitudinal chromatic aberration (axial)
transverse chromatic aberration (lateral)

## Image Sensors


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- Resolution
- Signal / Noise Ratio
- Cost


## Considerations

- Speed


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Fig. 4. Typical $512 \times 512$ CCD.

## Image Sensors

- Convert light into an electric charge


CCD (charge coupled device)
Higher dynamic range
High uniformity
Lower noise


INCIDENT LIGHT


CMUS (complementary metal
Oxide semiconductor)
Lower voltage
Higher speed
Lower system complexity

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## CCD Performance Characteristics

- Linearity Principle: Incoming photon flux vs. Output Signal
- Sometimes cameras are made non-linear on purpose.
- Calibration must be done (using reflectance charts)---covered later
- Dark Current Noise: Non-zero output signal when incoming light is zero
- Sensitivity: Minimum detectable signal produced by camera


## Sensing Brightness

Incoming light has a spectral distribution $p(\lambda)$


So the pixel intensity becomes

$$
I=k \int_{-\infty}^{\infty} q(\lambda) p(\lambda) d \lambda
$$

## How do we sense colour?

- Do we have infinite number of filters?


Three filters of different spectral responses

## Sensing Colour

- Tristimulus (trichromatic) values $\left(I_{R}, I_{G}, I_{B}\right)$

Camera's spectral response functions: $h_{R}(\lambda), h_{G}(\lambda), h_{B}(\lambda)$


$$
\begin{aligned}
& I_{R}=k \int_{-\infty}^{\infty} h_{R}(\lambda) p(\lambda) d \lambda \\
& I_{G}=k \int_{-\infty}^{\infty} h_{G}(\lambda) p(\lambda) d \lambda \\
& I_{B}=k \int_{-\infty}^{\infty} h_{B}(\lambda) p(\lambda) d \lambda
\end{aligned}
$$

## Sensing Colour



Bayer pattern
Foveon X3 ${ }^{\text {TM }}$

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## Resources

- J.C. Russ - Chapters 1 and 2
- L. Shapiro, and G. Stockman - Chapter 1
- "Color Vision: One of Nature's Wonders" in http://www.diycalculator.com/spcvision.shtml

