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Computer Vision

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Course Information

Course Information

- What the course is about
 - Computer Vision from Algorithms to Deep Learning
 - Starting with history and background information
 - Practical oriented course
- Timeplan
 - 15 Weeks, but there are also some holidays
 - Exception: lecture on 17.04
 - In week 3 (23.04. – 25.04), there will be no lecture (Lab retreat)

Course Information

- Homework
 - Four assignments in total
 - You can start them in the practical blocks on thursdays and ask questions
 - Each assignment takes 2–3 weeks
- Journal Club
 - Discussions of given papers (in groups)
- Exam
 - Final exam will be a group project with presentation and documentation
- Grading
 - Project presentation and documentation
 - Participation in journal clubs

Used Books

In the beginning of some chapters you will find a list of book sections relevant to the unit.
All abbreviations are explained here:

Main resources

FP: Forsyth and Ponce. Computer vision: a modern approach. 2nd edition.

SR: Richard Szeliski. Computer Vision: Algorithms and Applications. 2nd edition

GW: Gonzalez-Woods. Digital image processing. 4th edition

Supplemental material

DL: <https://www.deeplearningbook.org/>

SL: A Cookbook of Self-Supervised Learning

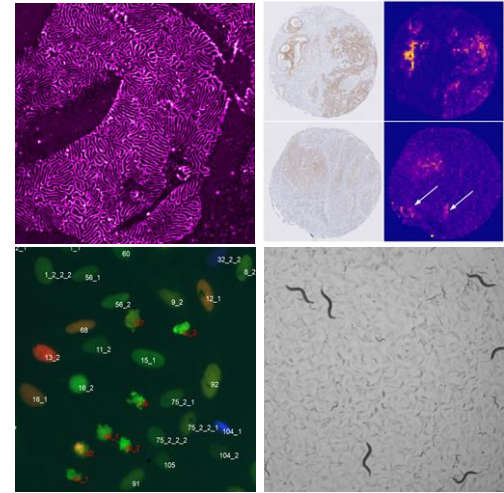
GP: Bichot, Siarry. Graph Partitioning

DIP: Burger and Burge. Digital Image Processing

DV: David Vernon: Machine Vision. Automated Visual Inspection and Robot Vision

Who are we...

- me: **Kasia Bozek**
- Professor for Data-Analytics in Bioinformatics
- Institute of Biomedical Informatics, Faculty of Medicine
- **Bozek Lab**
 - Develop and apply deep learning (DL) methods
 - Biomedical image data
 - cancer histopathology, microscopy images of podocytes, c. elegans behavior analysis, cell death prediction, ...
 - <https://bozeklab.com/>



Teaching Assistants (TAs)

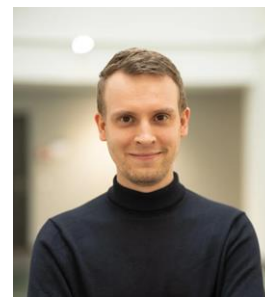
- PhD students from Bozek Lab
- Will do the practical sessions, correct the homework, ...
- If you have questions regarding the course, homework, etc.
contact: maurice.deserno@uni-koeln.de
- Put **CV-Course** at the beginning of the email subject



Florian



Juan



Maurice



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Introduction

Material:

- SR 1.1 – 1.2, 2.3.2
- GW 1.3, 2.1 – 2.2
- FP 1.1 – 2, 3.1–4

The power of human vision 🌈

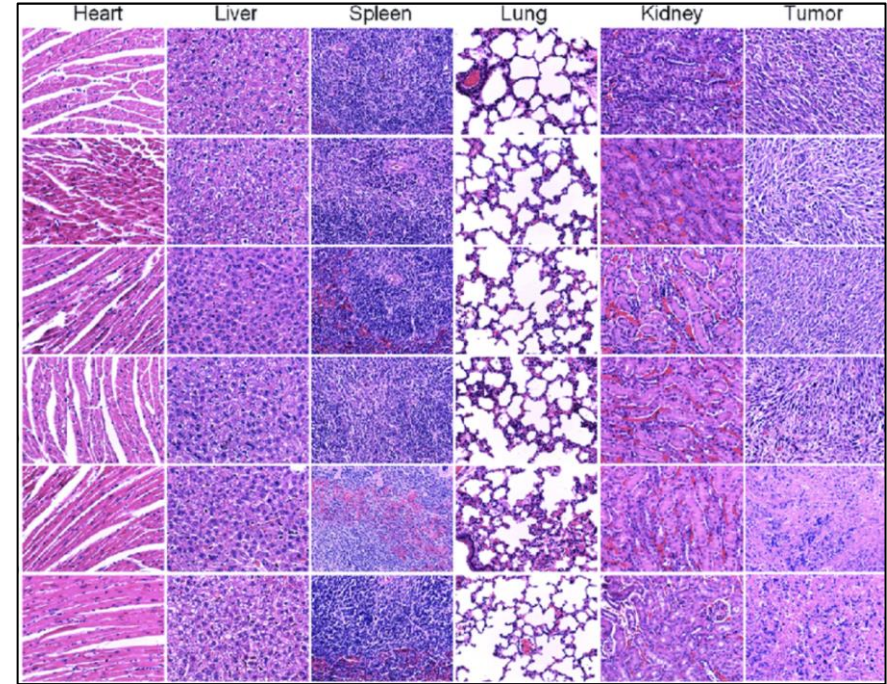
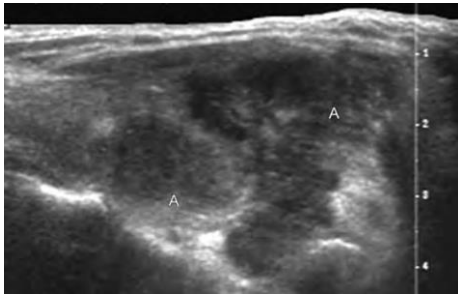
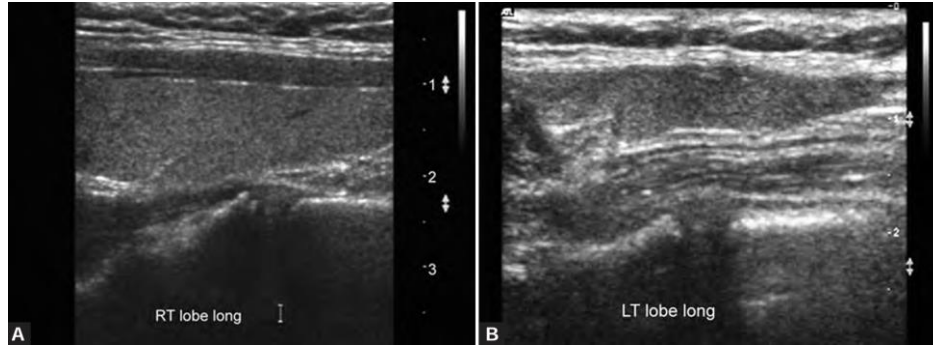
- What do you see?
 - A **bent** plant and some pots
 - A **fluffy** chair with a pillow
 - **Letters**, even a word you can read
 - Carpets on a **wood(-like)** floor
 - Located in a 3D room
 - ...
- All this you can extract from a 2D-RGB image 🦊
- How do you know all this?
- How to program a computer to “see” all this?



The power of human vision 🌈



The power of human vision 🌈



<https://www.aijoc.com/doi/AIJOC/pdf/10.5005/jp-journals-10003-1148>

<https://pubs.acs.org/doi/10.1021/acs.nanolett.6b02786>



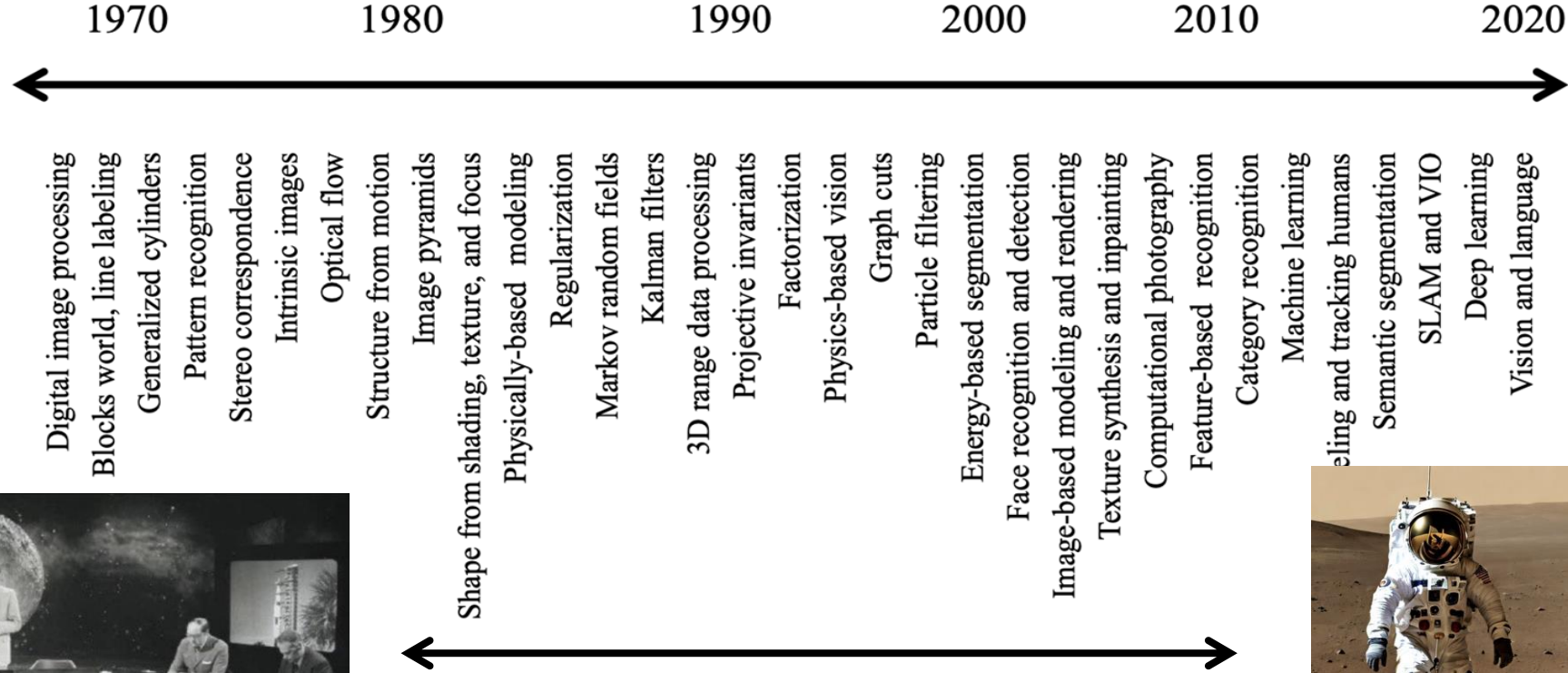
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Computer Vision

What is Computer Vision?

- Meaning changed over time
- Mimicking human vision in the 1960s – Today the field is quite broad
- Short: Enable a computer to see
- Include methods for:
 - acquiring
 - processing
 - analyzing and understanding digital images
 - extraction of high-dimensional data
 - produce numerical or symbolic information

What is Computer Vision?



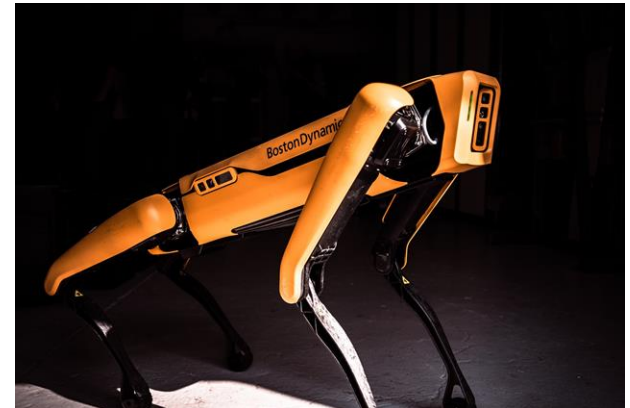
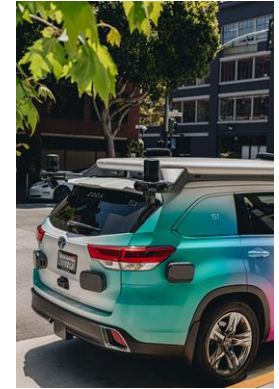
<https://www.scienceandmediamuseum.org.uk/objects-and-stories/moon-to-living-room-apollo-11-broadcast>



<https://huggingface.co/spaces/stabilityai/stable-diffusion>

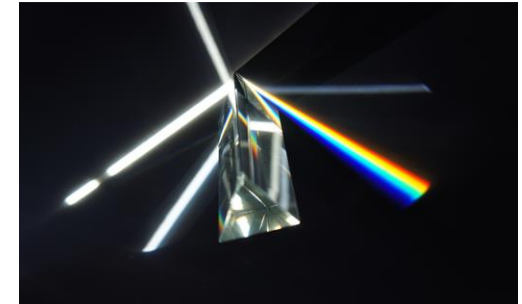
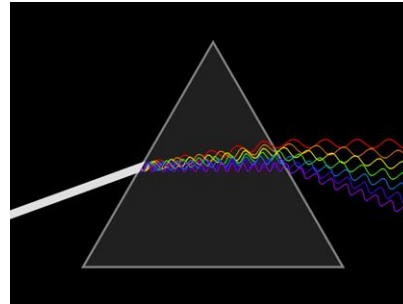
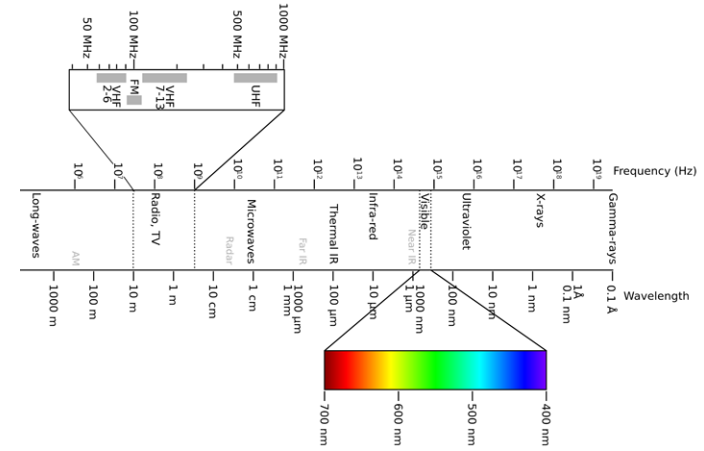
Applications 🤖

- Self-driving vehicles
- Robotics
- Bio-Medical Imaging
- Optical Character Recognition
- Retail/Warehouse-logistics
- Image/Video generation
- Surveillance
- ... many more



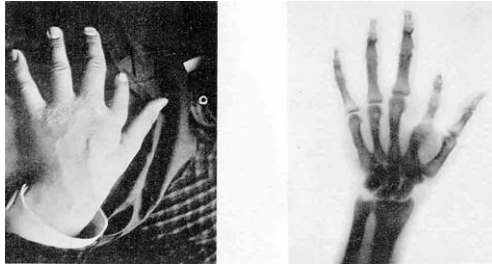
Electromagnetic Spectrum ⚡

- Isaac Newton first used the term **Spectrum**
- Principal energy source for images (nowadays)
- Alternatives sources: sound (eg. ultrasonic), synthetic data, ...
- Most known are images from the **human visible range**
- But other ranges are also used (see next slide)
- Imagine it as sinusoidal waves with a specific wavelength λ



Dispersive Prism
"splitting" white light

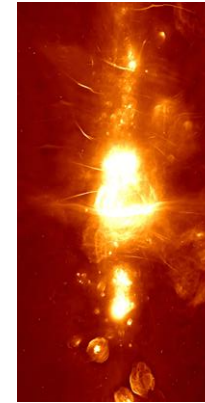
Some examples



X-ray of a hand



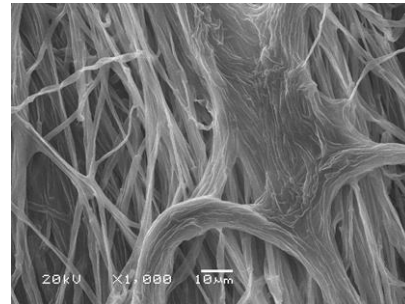
Different infrared ranges



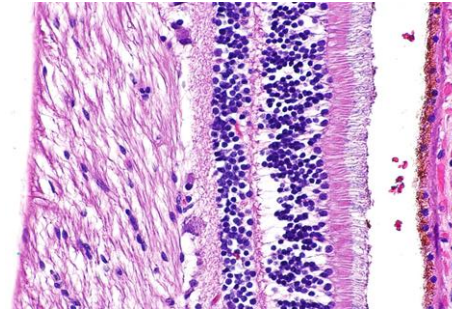
Radio image of the Milky Way



Ultrasonic image of a healthy 18 weeks old female



Mycelial film under a scanning electron microscope



Part of the retina under a microscope with H&E staining

Radio
Infrared
Microwave
Visible
Ultraviolet
X-ray
Gamma-ray



AM radio
Amateur radio
Aircraft communication
Microwave oven
TV Remote Control
Night vision goggles
Visible
UV light from the Sun
Airport security scanner
PET scan
Terrestrial gamma-ray flashes

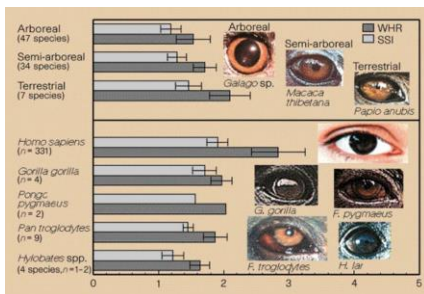
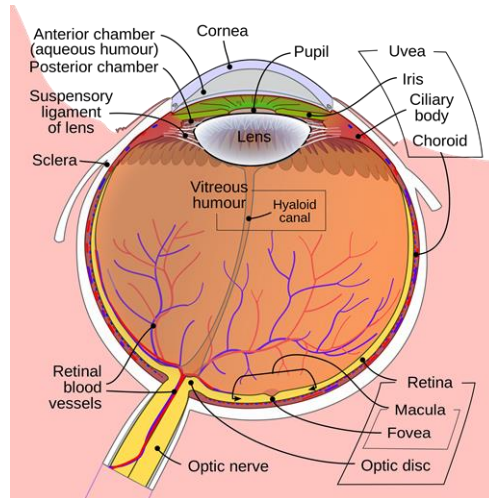


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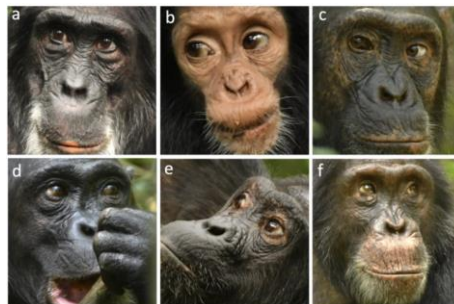
Computer Vision

Human Eyes 🧐

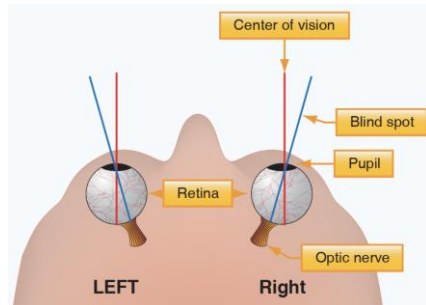
- Powerful sensory organ
- Only covered roughly
- Important parts of the eye:
 - Pupil: Controlling the amount of light hitting the lens
 - Cornea + Lens: Refracting the light, forming an image on the retina
 - Retina: Contains the photoreceptors, reacting to the light
- Shape of the lens varies to achieve a sharp image (change of focal length)
- Fun fact for CS people:
 - Human retinas bandwidth approx. 8.75 Mbit/s
 - Guinea pig's bandwidth approx. 875 Kbit/s



Kobayashi, H., Kohshima, S. 1997

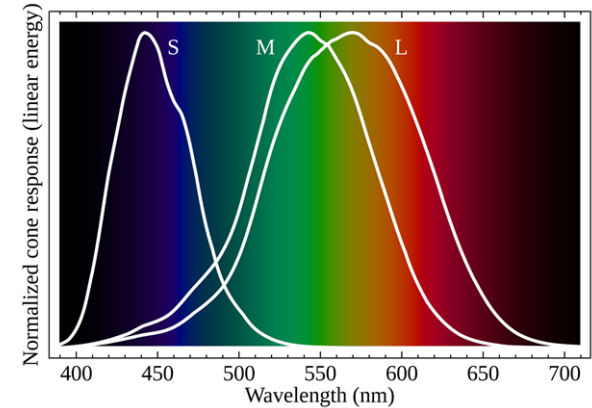


Clark et al. 2023



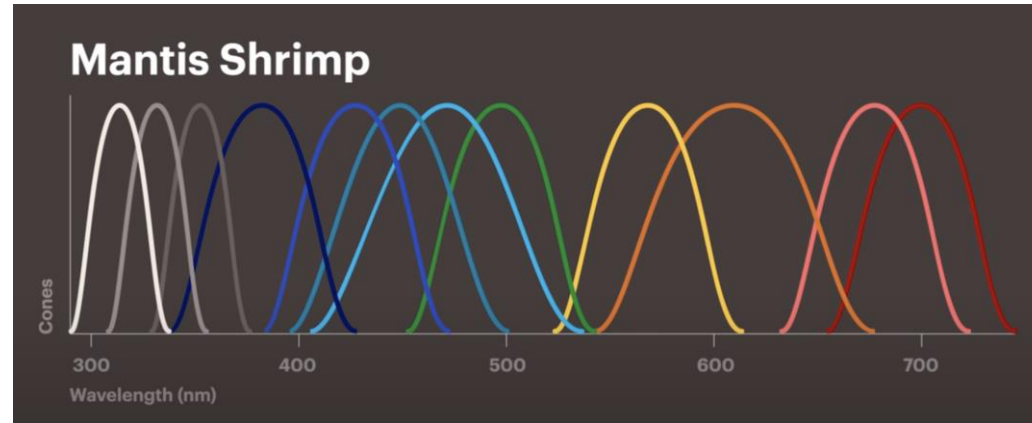
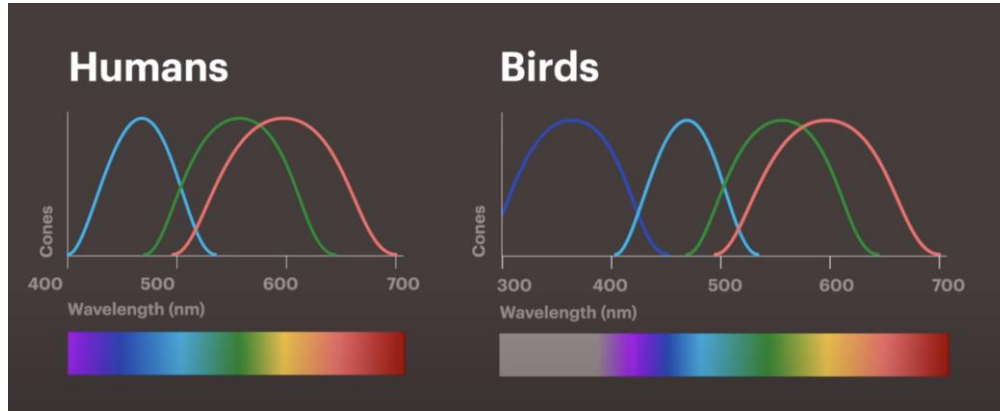
https://en.wikipedia.org/wiki/Cephalopod_eye

Light and Color in Human Vision



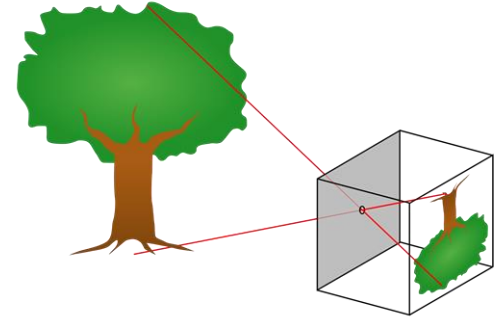
- Retina contains the photoreceptors
 - Rods and cones
 - Approx. 100 million rods, 5 million cones
- Rods are extremely sensitive to light and are important for night vision
- They play only a minor role in color vision (that's why you see less colors in darkness)
- Their density increases to the outer edges of the retina
- Cones are less light sensitive but enable color vision
- Three different types (S, M, L), categorized by their sensitivity to different wavelength (see image)

Light and Color in Human Vision



How to mimic an eye? – The Pinhole Camera

- Very basic camera – Uses no lenses
- Usually a **light proof box** with a **tiny hole** at one side
- “**Pinhole**” because the hole is commonly punched with a **pin**
- The small hole blocks rays which do not reflect directly from the object to the camera
 - Otherwise we would not achieve a clear image of the subject
- The smaller the hole, the sharper the image (but also darker)
- Image appears upside down (like inside the human eye)
- One thing is missing to capture the actual image... The retina!
- The image can be captured analog by a film... or by a digital **sensor**!

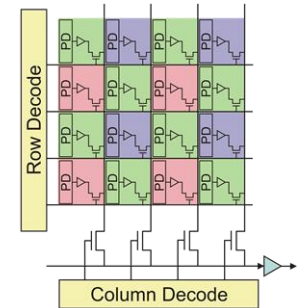


From Human Vision to Computer Vision 🧑💻

- Most practical method: Digital photography, recording a color image
- Instead of your **rods** and **cones** an image sensor uses **photodiodes** (PD)
- Photodiodes produce electricity when hit by photons (e.g. from visible light)
- To detect colors (**specific wavelength**) a color filter array is placed over the sensor
- Common filter is the **Bayer Filter** consisting of $\frac{1}{2}$ green, $\frac{1}{4}$ red and $\frac{1}{4}$ blue filters



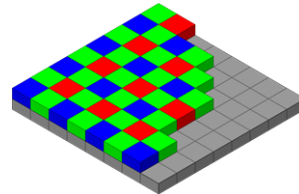
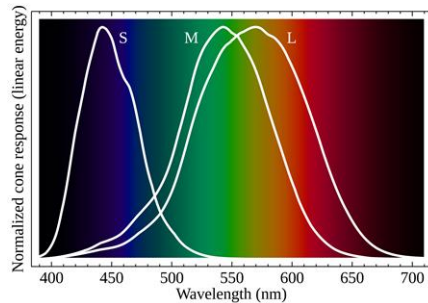
CMOS camera sensor



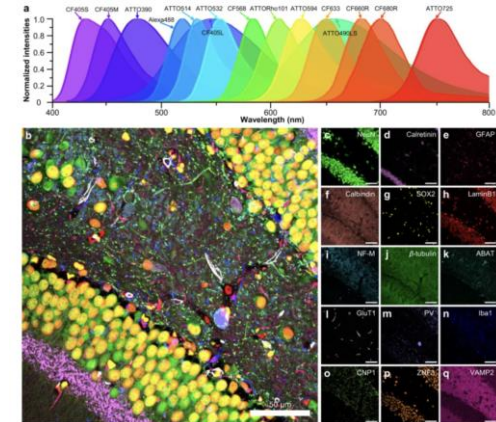
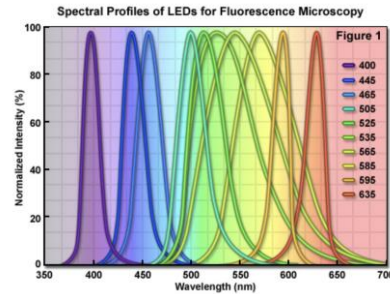
Photodiodes (PD) of a CMOS Image Sensor with Bayer Filter

Enabling digital color vision – Bayer Filter

- The bayer filter is used so that a photodiode “reacts” only to a specific color
- Otherwise we would not know which color (wavelength) the photon had
- But why does the bayer filter have $\frac{1}{2}$ green pixels?
- To mimic the **human perception** of color!
 - Our cones are most sensitive to green light (see graph)
- To get a red, green and blue (RGB) value for each square (pixel) demosaicing (interpolation) is applied

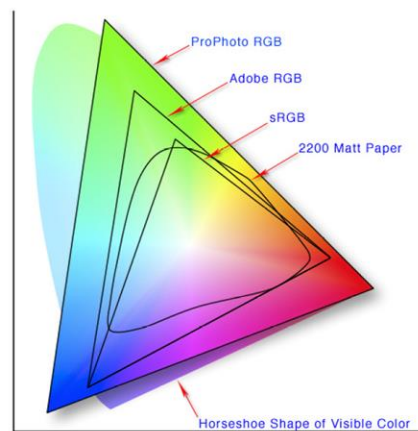


Bayer Filter Array



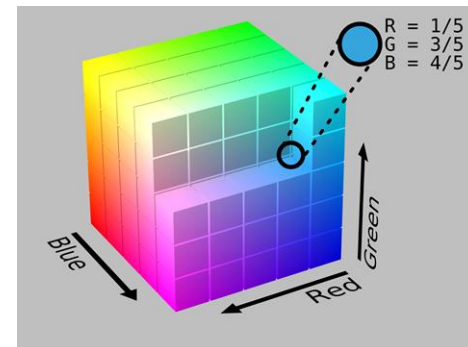
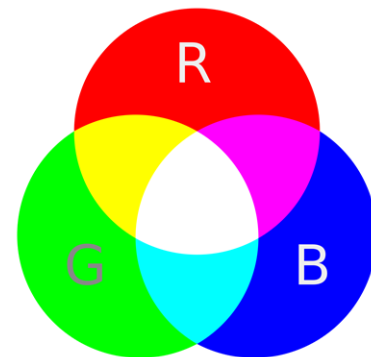
Color Spaces

- We already talked about RGB, let us define this a bit more
- A **color space** is a combination of a **color model** and a **mapping function**
- A **color model** describes how colors can be represented as numbers
- Examples of **color models**:
 - RGB: Additive color mixing; individual values for the three channels red, green and blue
 - CMYK: Subtractive color mixing; individual values for the four channels cyan, magenta, yellow and black
 - HSV: encodes color with individual values for hue, saturation and value
- Common **color spaces** are: sRGB, Adobe RGB or DCI-P3; all using RGB color model but covering different colors (see image)



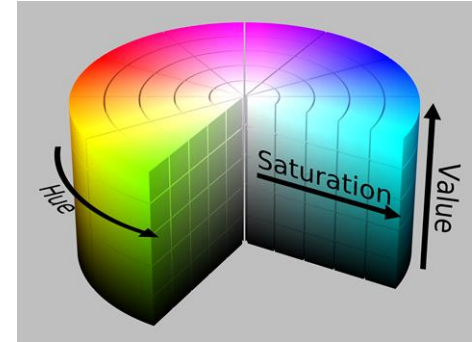
RGB – Color Model 🦸

- Probably the most common one
- Describes which light needs to be emitted for a specific color
 - Additive Model → Combine red, green and blue to get white color
 - No color (0, 0, 0) → Black image
- RGB color spaces can be mapped to a cube
- Commonly used for images in computer vision
- Typically colors of a pixel are identified by a triplet (R, G, B)
 - Floats between 0 and 1 → (0.96, 0.62, 0.11)
 - Integer between 0 and 255 (8 bit) → (245, 157, 29)
 - Hexcode → #F59D1D
 - If all three values are the same, we get a grayscale color



HSV – Color Model

- Cylindrical coordinate representation in a RGB color model
- Defines a color with three values:
 - Hue: **angular axis**, encodes the base color
 - 0° red, 120° green, 240° blue
 - Saturation: defines the intensity of the color
 - Value (also called brightness): encodes the perceived brightness
- Often used in CV
 - Many CV algorithms for color images are extensions of the grayscale version
 - As a result, sometimes each color component is passed individually
 - With RGB all channels are correlated (all encode the amount of light hitting the object)
 - With HSV they are easier to separate



Looking at bits

- Until now we talked about the hardware to capture an image and how color is represented
- But how does an image look like “under the hood”?
- A simple and common way is to store an image as a **matrix of pixel values**
- The image coordinate systems origin (0, 0) lies in the **top left corner**
- For grayscale images a pixel is represented by **one value**, resulting in a 2D array
- For RGB images a pixel is represented by **three values**, resulting in a 3D array
- In practice images often come with a lot of (meta-)information
 - In photography: GPS coordinates, camera model, aperture f-number, ...
 - In bioimages (e.g. microscopy): magnification, resolution, data to identify the sample/experiment, ...
- Images can also be 2-byte, 12-bit and contain many channels!

