



# CAP 4453 Robot Vision

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#### Course Website + resources

- CANVAS (webcourses)
- Alternative: CAP 4453 Fall 2022 (ucf.edu)
- Colab notes:gonzo1978/CAP4453: Colab notes for CAP 4453 (github.com)
- Szeliski, Computer Vision: Algorithms and Applications
- <u>Python for Computer Vision</u>. A tutorial will be given in the class on PyTorch for deep learning.



### **Course** logistics

- Class time: Tuesday and Thursday 3:00pm 4:15pm
  - Classroom: CB2 O105
- Office hours [Zoom]
  - Tuesday 7:00pm-8:00 pm
  - By appointment (send me an email)



#### Course Outline

- Image Filtering
- Edge Detection
- Feature Extraction
- Image warping
- Optical Flow
- Basics of Neural Networks
- Deep Learning for Computer Vision
- Image Segmentation
- Image Classification
- Object Detection



## Grading

- Weekly homework: 25%
- Mid term exam: 25%
- Final exam: 30%
- Programming project 1: 10%
- Programming project 2: 10%



#### Scores

•95-100 = A
•90-94 = A•85-89 = B+
•80-84 = B
•75-79 = B•70-74 = C+
•65-69 = C
•60-64 = C•55-59 = D+
•50-54 = D
•45-50 = D•0-44 = F



### Grading criteria

- Programming: homework must be written in python. They should be written as colab notes. They should include explanatory/clear comments as well as a short report describing the approach, detailed analysis, and discussion/conclusion. Optionally, you could add GUI to your projects.
- Collaboration: Students are free to discuss ideas and technical concepts. However, students must submit original work for all assignments, projects and exams, and abide by UCF Golden Rule. Cheaters will not be tolerated.



## Questions?





# Robot Vision

I. Introduction



#### Robot vision

- Ability of robots to understand visual data from the world using:
  - Hardware: like 2D cameras, 3D stereo cameras
  - Computer algorithms
- Goal: automate task which human visual system can perform









## Vision and Image Understanding

- Visual tasks: We use vision to interact with environments and survive
  - to navigate and avoid obstacles
  - to recognize and pick up objects
  - to identify food and danger
  - ... friends and enemies
  - ...

### Visual Perception (6 blind man and an elephant)



# We need a Visual sensor





#### Robot vision Vs human vision





#### **Visual Perception**

 Definition: Process of acquiring knowledge about environmental objects and events by extracting information from the light they emit or reflect [Palmer, 2012].





Retina

Perception is analogous to taking a picture! (credit: Palmer, 2012)



#### A large list of visual sensors



Monochrome cameras



**RGB** cameras





Lidar / Time of flight cameras  $_{14}$ 



### What is a (digital) image?



From imagilabs.com



## Goal of Robot Vision?

- To bridge the gap between
  - image pixels and "meaning" (semantic)!



#### What we see!





#### What robot sees!



### Capturing a color image

Understanding Digital Camera Sensors (cambridgeincolour.com)



CCD Bayer sensor





## Capturing a color image









## Capturing a color image

Understanding Digital Camera Sensors (cambridgeincolour.com)



Next steps:

**Demosaicing artifacts** 

N<sub>cols</sub>

- Optical low pass filter -
- Anti-aliasing filtering -

Left hand coordinate system



#### Anti-Aliasing





#### Typical CCD sensor sizes



- Industrial cameras usually use 1/3" sensors in case of resolutions of 640 x 480 pixels, cameras with 1280 x 1024 pixels mainly 1/2".
- The quite popular camera resolution of 1600 x 1200 pixels often uses a somewhat larger sensor with 1/1.8" with the same pixel size.
- Sensors of consumer cameras (8 to 12 megapixels for 200 euros) have pixel sizes of mostly 1.7 µm today
- Machine vision cameras (C-mount) with resolutions from VGA to 2 megapixels normally have pixels of 4.6 to 6.5 µm with a 10 15 times larger light-active surfaces and thus clearly better signal results. If you need images as noise-free as possible and precise measuring results, look for preferably large sensor pixels, even if these cameras are more expensive!
- Well capacity: This specification describes how many electrons a pixel element can hold before it is completely saturated. A pixel of 5.5 µm structure size can accumulate approximately 20,000 electrons, a 7.4 µm pixel 40,000 electrons.
- The larger the full well capacity, yet the better the maximum signal-noise ratio. Consumer cameras with pixel sizes of 1.7 μm require only about 1,000 photons for the pixel saturation.



## Image Type: RGB (red, green, blue)

- Image has three channels (bands)
- Each channel spans a-bit values.



Human Cone-cells (normalized) responsivity spectra



Some people might have 4 cone-types! Some might have just 2!



### Color

#### Color vision has evolved over millions of years.





#### Color

- If there is no light, there is no color!
- Human vision can only discriminate a few dozens of grey levels on a screen, but hundreds of thousands of different colors.
  - RED -> ~625 to 780 nm
  - ORANGE -> ~ 590 to 625 nm
  - YELLOW -> ~565 to 590 nm
  - GREEN -> ~ 500 to 565 nm
  - CYAN -> ~485 to 500 nm
  - BLUE -> ~440 to 485 nm
  - VIOLET -> ~330 to 440 nm

[long wavelength] [long wavelength] [middle range wavelength] [middle range wavelength] [middle range wavelength] [short wavelength] [very short wavelength]



#### STOP CENTRY E 63 + 400

## Retina of Human Eye



There are three different types of colorsensitive cones corresponding to (roughly)

- RED (64% of the cones)
- GREEN (about 32%), and
- BLUE (about 2%).

6-7 million cones 120 million rods

Credit: Klette, 2012.



#### Demo: Color is in your head !





#### A large list of visual sensors



Monochrome cameras



**RGB** cameras





Lidar / Time of flight cameras



### Multispectral cameras









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#### MultiSpectral cameras



<u>Multispectral Imaging: New Technology Resurrects</u> <u>Centuries-Old Texts (nbcnews.com)</u>

- PCB inspection
- Skin characterization
- Food inspection
- Agriculture
  - Analyzing crops
- Military



### Computer Vision 2010 vs Today





In 2010

### Applications

- Self Driving cars
- Biometric verification
- Healthcare (Gauss Surgical)
- Medical diagnostics
- Retail (amazon go)
- Industrial monitoring
- Agriculture (remove weeds)

#### Cargill brings facial recognition capability to farmers through strategic equity investment in Cainthus

Sensors give farmers clear picture of animal health and well-being

(MINNEAPOLIS) January 31, 2018 — Cargill and Cainthus, a Dublin-based machine vision company, are reshaping how animal producers make decisions for their livestock through a strategic partnership that will bring facial recognition technology to dairy farms acrose the work). The deal includes a minority equity investment from Cargill. Terms were not disclosed.

Cainthus uses breakthrough predictive imaging to monitor the health and wellbeing of livestock. Their proprietary software uses images to identify individual animals based on hide patterns and facial recognition, and tracks key data such as food and water intake, heat detection and behavior patterns The software then delivers analytics that drive on-farm decisions that can impact milk production, reproduction management and overall animal health.



"We are enthused about what this partnership will mean for farmers across the world," said David Hunt, president and co-founder, Cainthus. "Cargill is a natural partner for us, given their focus on bringing a world-class digital

capability to the market and their understanding of how technology will truly help farmers succeed. We think this partnershi will be a game changer for farmers because it will allow them to efficiently scale their business."



Google's DeepMind Beats Doctors at Detecting 50 Eye Diseases Just by Looking at Scans

By KC Cheung Last Modified Date - May 27, 2020





#### The Machine Making sense of Al

#### AI startup Gather uses drones and computer vision for warehouse inventory

Iohnson @kharijohnson August 15, 2019 12:10 PM



Al startup Gather uses drones and computer vision for warehouse inventory | VentureBeat

#### Forbes

#### Walmart Expands Use of Bossa Nova's Robots from 50 to 350 Stores



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Walmart Expands Use of Bossa Nova's Robots from 50 to 350 Stores (forbes.com)

() This article is more than 2 years old.



Bossa Nova Robotics at Being Used at Walmart Stores BossA NovA ROBOTICIS

It was recently announced that Walmart has expanded its use of autonomous mobile robots (AMRs) used for real-time, on-shelf product data. Walmart will put these AMRs in 350 stores. The robots come from Bossa Nova Robotics.

#### Disney's facial recognition AI watches you watch movies | Daily Mail Online

#### Disney reveals AI that can monitor cinemagoer's faces to reveal what they REALLY thought of the movie

- Disney is using neural networks to learn about stereotypical reactions to films
- After observing viewers for only a few minutes, it can predict facial expression for the remainder of the movie
- The system 'taught' itself to recognize smiling and laughing all on its own

By SAGE LAZZARO FOR DAILYMAIL.COM PUBLISHED: 17:58 EDT, 24 July 2017 | UPDATED: 11:10 EDT, 25 July 2017



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### What changed?

- Emergence of deep learning
- Advancement in hardware
- Availability of large-scale data
  - ImageNet
  - OpenImages
  - YFCC100M
  - Youtube-8M
  - Kinetics
  - AVA





#### Hardware



#### Train

#### CPU

Small models

Small datasets

• Useful for design space exploration

#### GPU

- Medium-to-large models, datasets
- Image, video processing
- Application on CUDA or OpenCL

#### TPU

- Matrix computations
- Dense vector processing
- No custom TensorFlow operations



#### FPGA

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- Large datasets, models
- Compute intensive applications
- High performance, high perf./cost ratio

#### Inference

Device	Pros	Cons
CPU	Cost effective, fit for general purpose, powerful cores, high memory capacity	Don't fully exploit parallelism, low throughput performance
GPU	High throughput performance, a good fit for modern architectures (ConvNets)	Expensive, energy-hungry, has IO latency, memory limitations
Custom Al Chips (ASIC, SoC)	Potential to significantly boost inference performance	Expensive and hard to develop
FPGA	Chip, energy efficient, flexible	Extremely difficult to use, not always better than CPU/GPU

How to Choose Hardware for Deep Learning Inference | Deci



## Questions?