Course: Deep Learning

**Unit 2: Computer Vision** 

# Introduction to Computer Vision

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### Introduction to Computer Vision

- 1. What is Computer Vision?
  - Related problems
  - Areas of interest
  - Applications
  - Deep Learnig in Computer Vision
- 2. Object recognition
  - Why is it difficult?
- 3. Miscelaneous
  - Overview of unit 2

• What is computer vision?

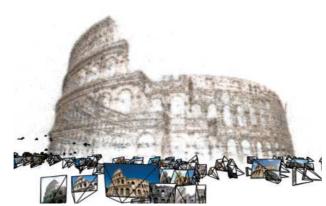
Make machines that "see"; obtain information from images.

Recognize places



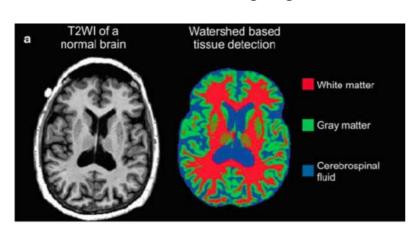
Look at people

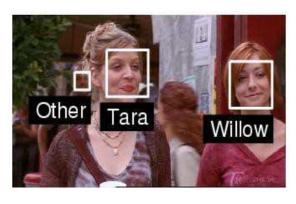


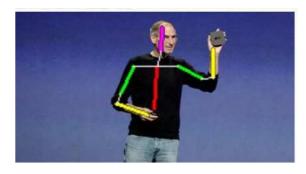




Medical imaging

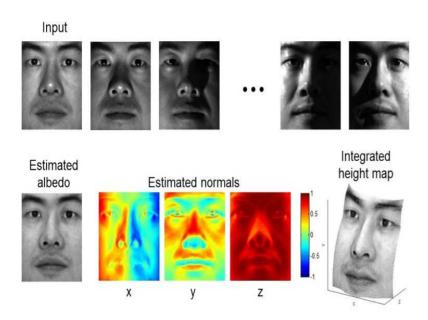


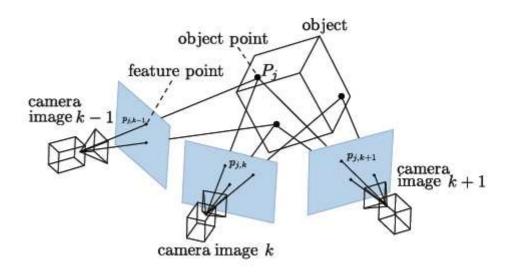




• What is computer vision?

- Reconstruccion
  - Shape from X (estéreo)
  - Shadows and shading
  - Shape fron defocus



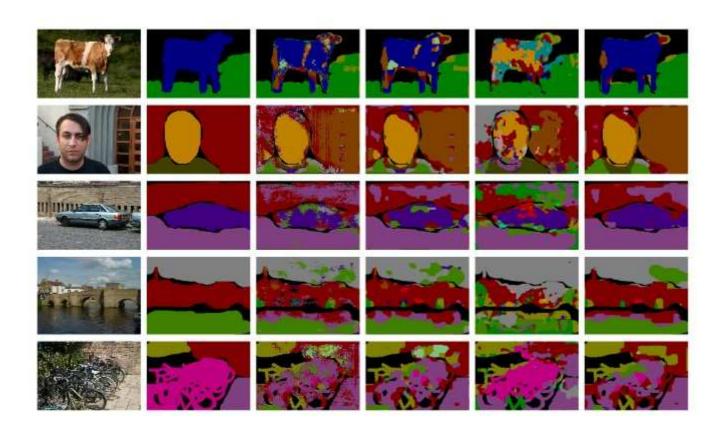






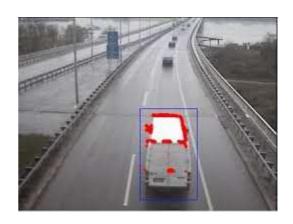
• What is computer vision?

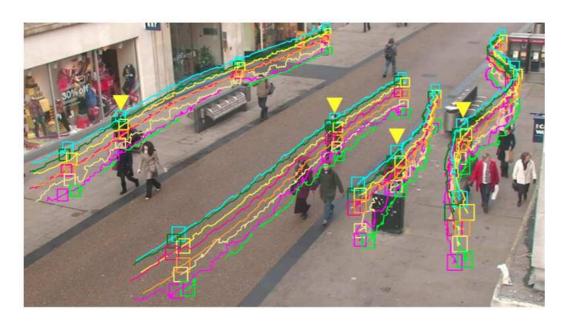
- Reconstruccion
- Segmentation



• What is computer vision?

- Reconstruccion
- SegmentationTracking





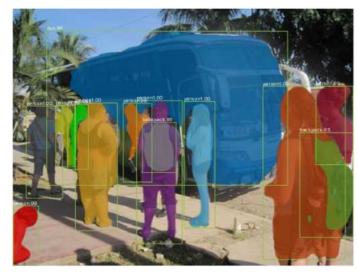


• What is computer vision?

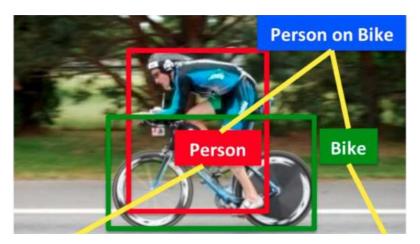
- Reconstruccion
- Segmentation
- Tracking
- Recognition



Image classification



Instance segmentation



Object detection/localization

• What is computer vision?

- Reconstruccion
- Segmentation
- Tracking
- Recognition
- Interpretation



"man in black shirt is playing guitar."



"construction worker in orange safety vest is



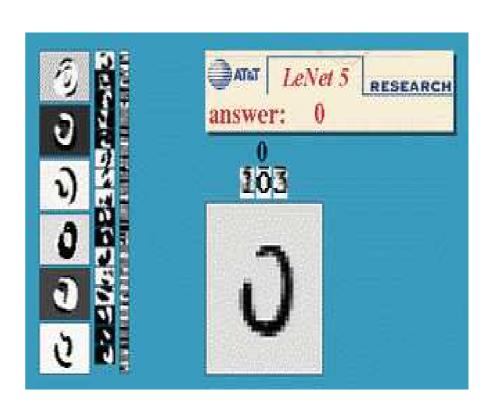


Applications

#### **Optical Character Recognition (OCR)**

Reading checks and forms





OCR, LeNet, LeCun 1989

Applications3D reconstruction. Hawk Eye



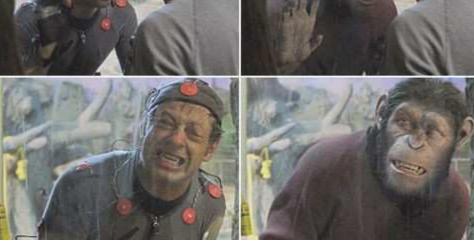
http://www.hawkeyeinnovations.co.uk/page/sports-officiating/tennis

Applications

**Movies** (graphics animation, augmented reality) Avatar, Rise of the planet of the apes, ....

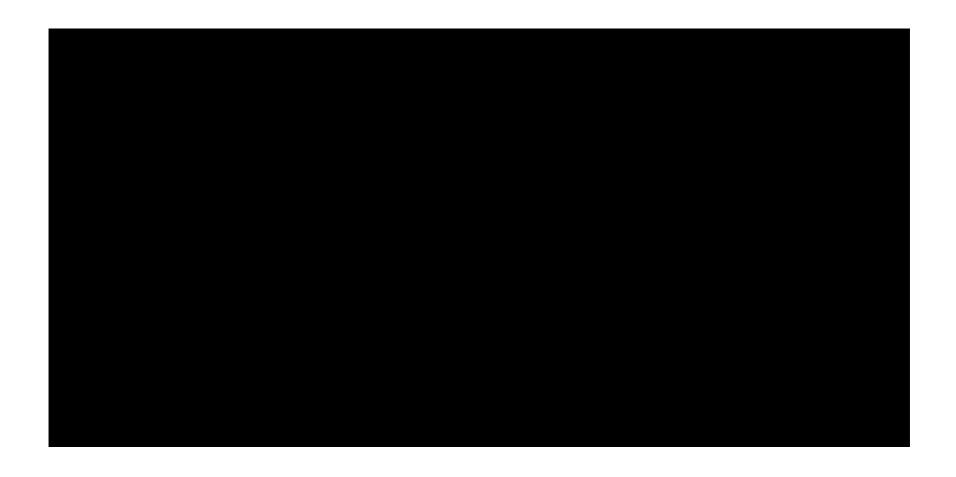






Applications

**Movies** (graphics animation, augmented reality) Avatar, Rise of the planet of the apes, ....



Applications

#### **Computational photography**



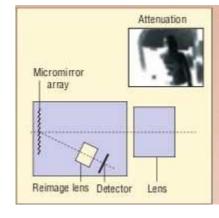














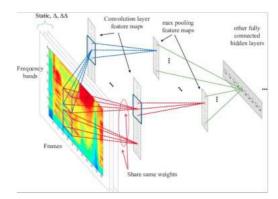




### Deep Learning and perception

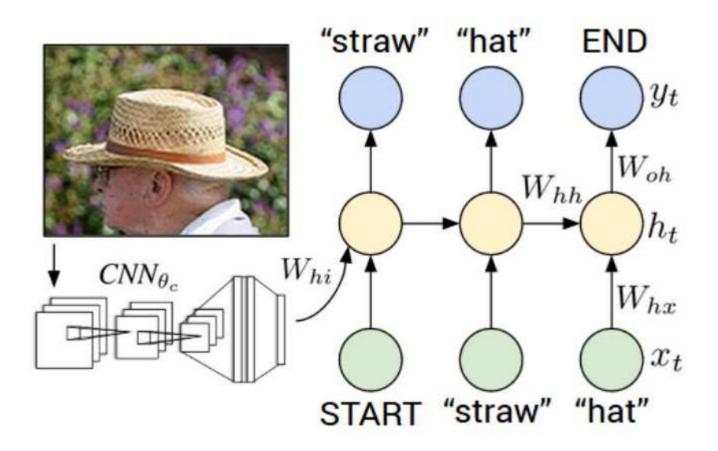
 In the past perception modalities used different technologies

- Deep learning is the state of the art for
  - Computer vision
  - Speech recognition
  - Natural language



- DL provides a common ground to address perception
- Opens the way for ultimodal communication

Deep Learning resultsImage to text



Karpathy, Deep Visual-Semantic Alignments for Generating Image Descriptions. CVPR, 2015.

Deep Learning resultsImage to text



"man in black shirt is playing guitar."



"construction worker in orange safety vest is working on road."



"two young girls are playing with lego toy."



"boy is doing backflip on wakeboard "

Karpathy, Deep Visual-Semantic Alignments for Generating Image Descriptions. CVPR2015.

## Deep Learning resultsText to image

this magnificent fellow is almost all black with a red crest, and white cheek patch.



this white and yellow flower have thin white petals and a round yellow stamen



Text descriptions Images (content) (style)

The bird has a **yellow breast** with **grey** features and a small beak.

This is a large white bird with black wings and a red head.

A small bird with a **black head and** wings and features grey wings.

This bird has a **white breast**, brown and white coloring on its head and wings, and a thin pointy beak.

A small bird with **white base** and **black stripes** throughout its belly, head, and feathers.

A small sized bird that has a cream belly and a short pointed bill.

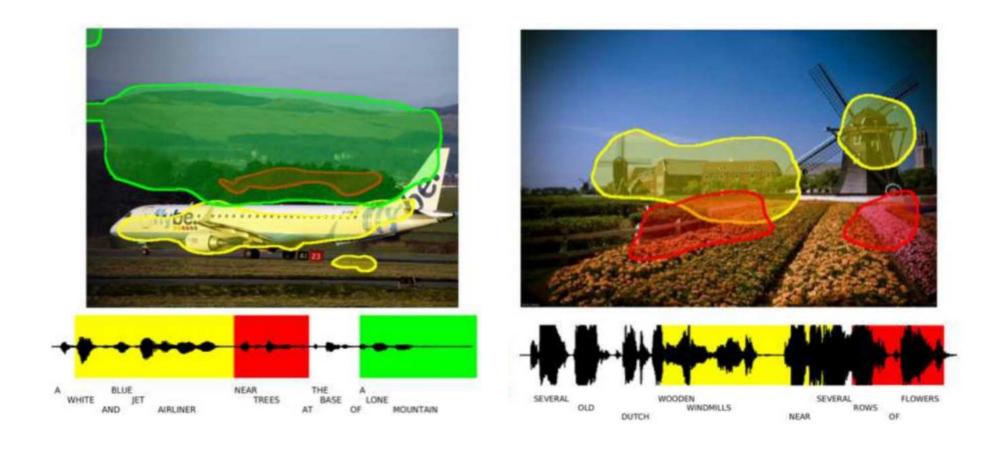
This bird is completely red.

This bird is completely white.



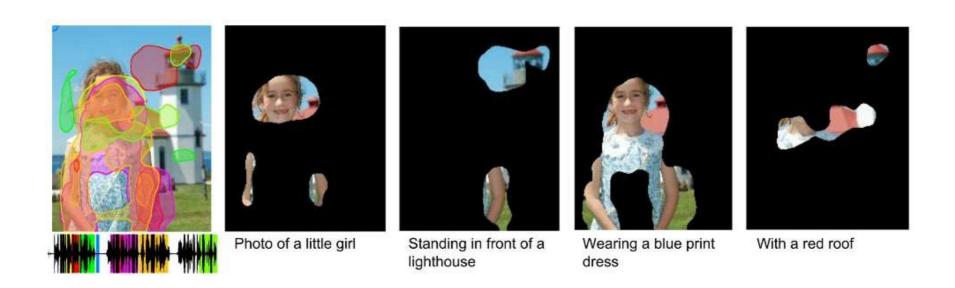


Deep Learning results
Self learning semantic correspondences between speech and pixels with no annotated data



Harwath, Jointly Discovering Visual Objects and Spoken Words from Raw Sensory Input. ECCV18.

Deep Learning results
Self learning semantic correspondences between speech and pixels with no annotated data



Deep Learning results.

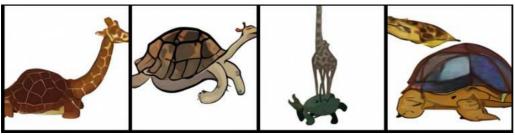
Image generation from text description, trained from 250M imagetext pairs from the Internet



A collection of glasses sitting on a table



An arcmchair in the shape of an avocado



A jiraffe made of turtle

Deep Learning applications



Automated retail



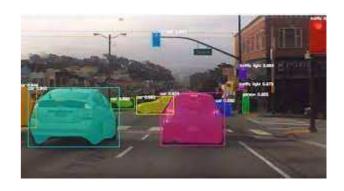
Augmented reality



Human pose



Face recognition



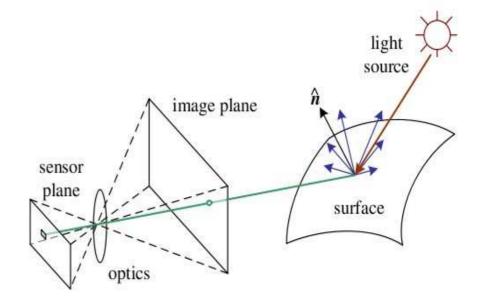
Self-driving

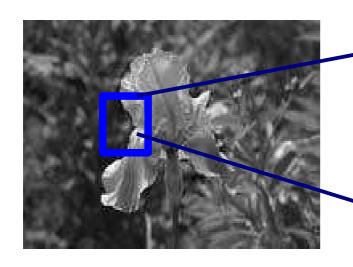


**Driver monitoring** 

• Why is vision hard?

It is an inverse problem





45	60	98	127	132	133	137	133
46	65	98	123	126	128	131	133
47	65	96	115	119	123	135	137
47	63	91	107	113	122	138	134
50	59	80	97	110	123	133	134
49	53	68	83	97	113	128	133
50	50	58	70	84	102	116	126
50	50	52	58	69	86	101	120

- Why is object recognition hard?High variability of natural object classes









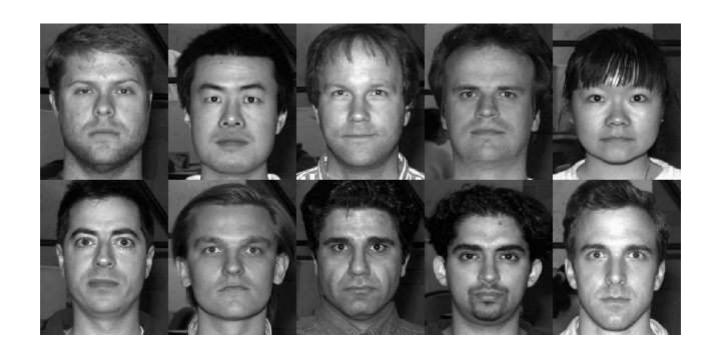








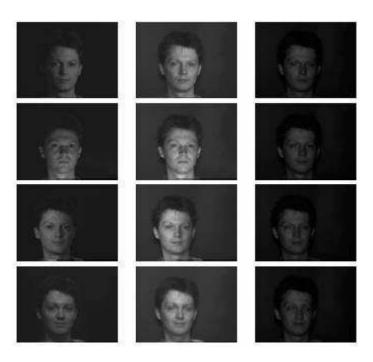
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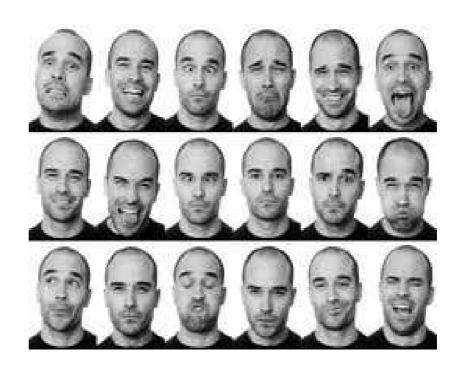
- Why is object recognition hard?
  - High variability of natural object classes
  - Lighting contrast, shadows, specularities



- Why is object recognition hard?High variability of natural object classes
  - Lighting contrast, shadows, specularities
  - Geometric variability



- Why is object recognition hard?
  - High variability of natural object classes
  - Lighting contrast, shadows, specularities
  - Geometric variability
  - Deformation



- Why is object recognition hard?
  - High variability of natural object classes
  - Lighting contrast, shadows, specularities
  - Geometric variability
  - Clutter, occlusion
  - Deformation

#### How is it addressed?

#### **Overview of Unit 2:**

- 1. Introduction to
  - computer vision,
  - image processing, convolution, object recognition,
- 2. Convolutional Neural Networks
- 3. Deep Architectures
  - AlexNet, VGG, LeNet, ResNet,...
  - MobileNet, Efficient Net, ...
- 4. Representation Learning
- 5. Computer Vision applications
  - Segmentation, Object Detection