

CAP 5415:

Computer Vision



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Computer Vision:

Various definitions. Some common definitions are:

- Study of visual data.
 - Examples for visual data: Images, Videos
- A scientific field that extracts meaning (information) from images or videos. Examples are:
 - Visual recognition,
 - Look for a particular and a local activity occurring in videos,
 - Track the objects over time in videos, etc.
- A field that builds algorithms to understand & extract the content of images (or videos) and to use it for other applications. Examples are:
 - Robotics,
 - Medical field,
 - Surveillance,
 - Entertainment, etc.

Logistics:

Exams:

- See Syllabus for details.
- 1 Midterm

Hands-on Experience:

- Final Project
- Programming assignments (PA)

Presentation Experience:

- Research paper presentation
- Final project presentation

Office hours:

- Tuesday/Thursday 4:30pm-5:30pm
 - (after the class)
- Office: HEC 214

Notes:

- Follow the slides
- Special thanks to:
- Dr. Mubarak Shah
 - Dr. Ulas Bagci
 - Dr. Yogesh Singh Rawat

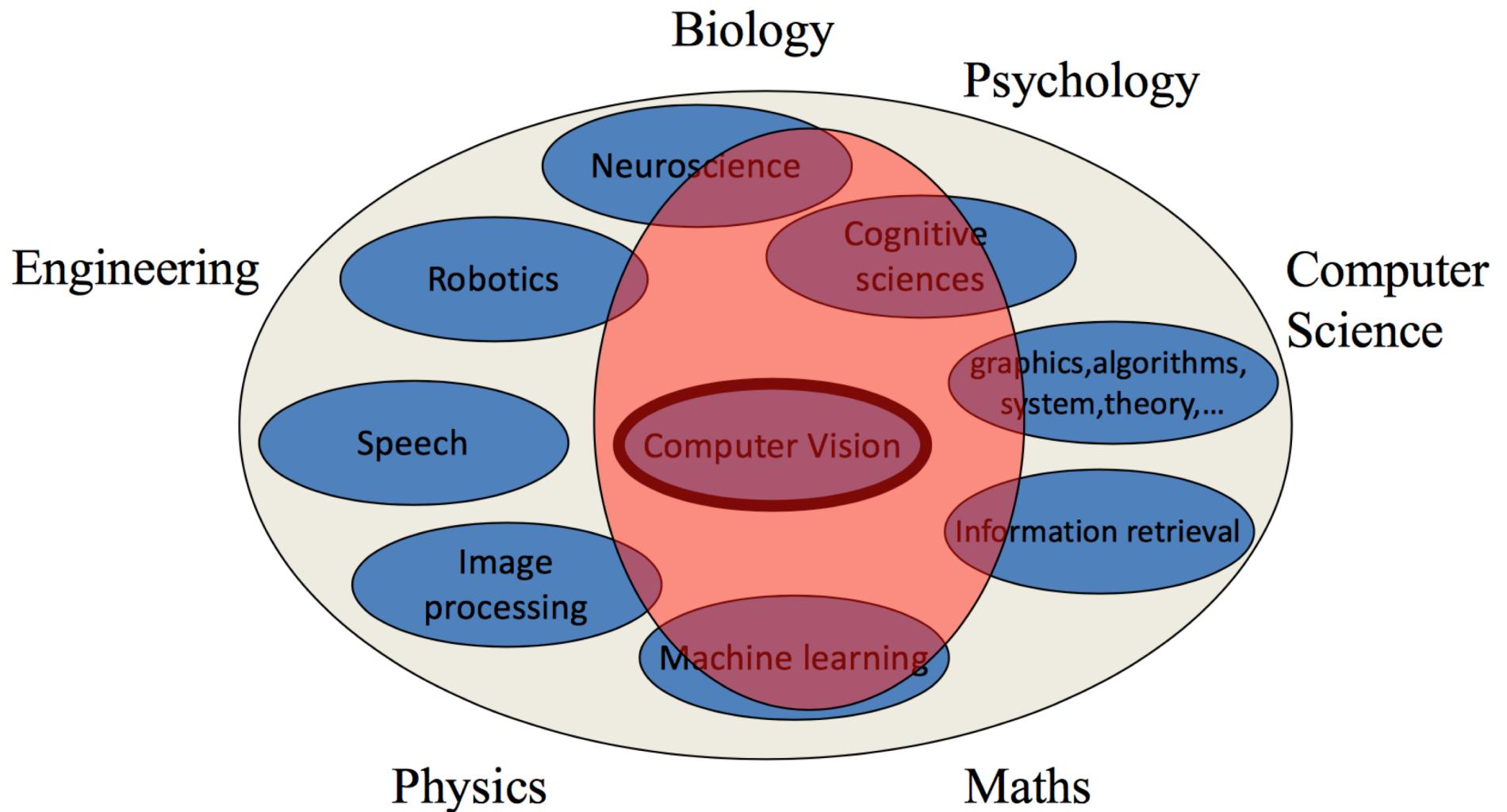
Teaching Assistants:

- Google
- Bing
- TBA

- No text book is required!
 - There are many online resources available.
- Attend the class (mandatory), follow the lecture notes and in-class material.
- Python is the required language for PA.

Computer Vision

What is it related to?



Computer vision: introduction



Visual recognition: Classification

Problem:

Are there any human in this image?

Answer: Yes / No

Problem:

Are there any cars in this image?

Answer: Yes / No

:



Visual recognition: Detection (localization)

Problem:

Tell me if there is any person in this image and if there is, where in image?

Answer: in the form of bounding boxes on the image.



Visual recognition: Object Categorization

Problem:

Detect all the objects in this image and give me a list of their type.

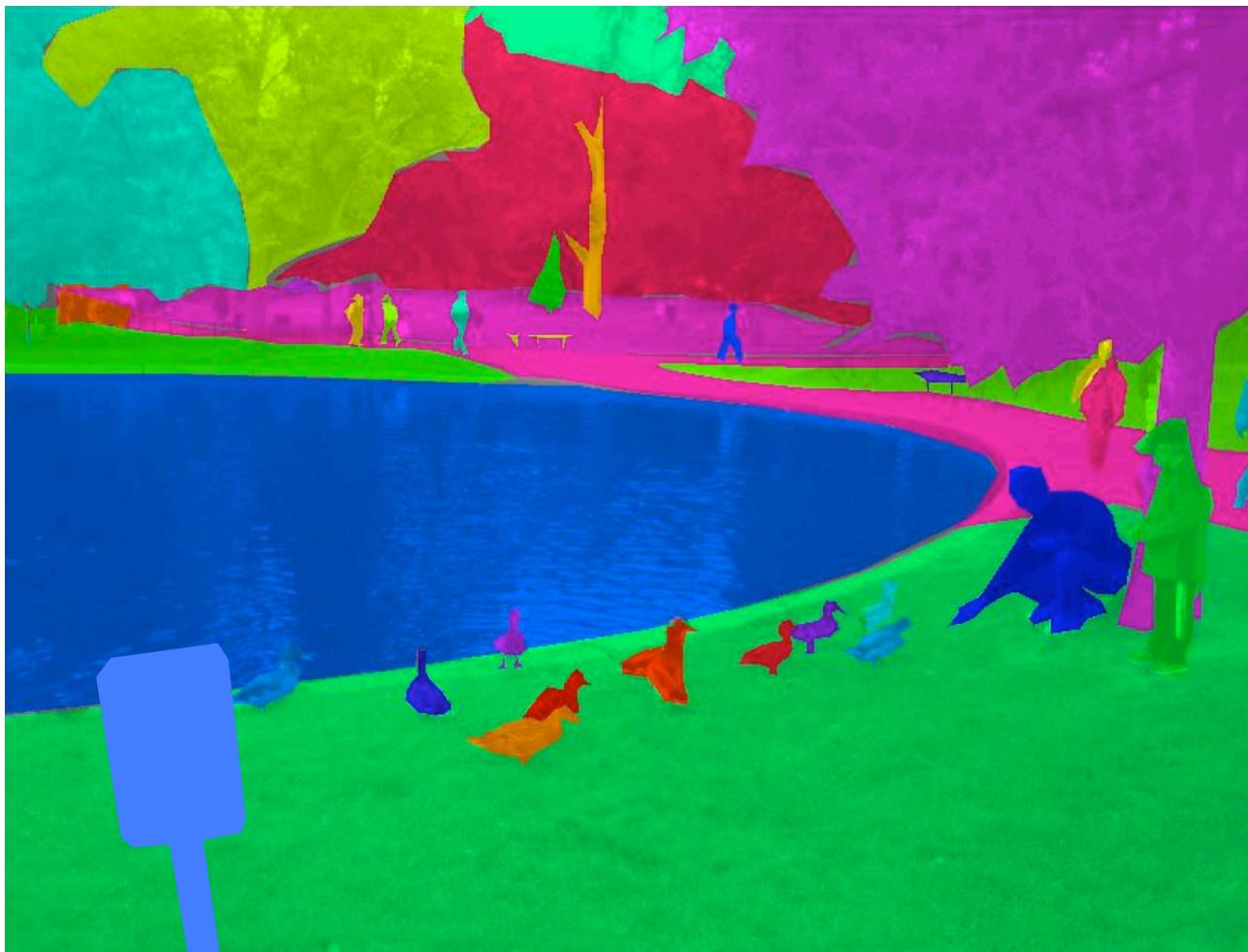
Answer: Building, Tree, Sky, People, Human, Grass



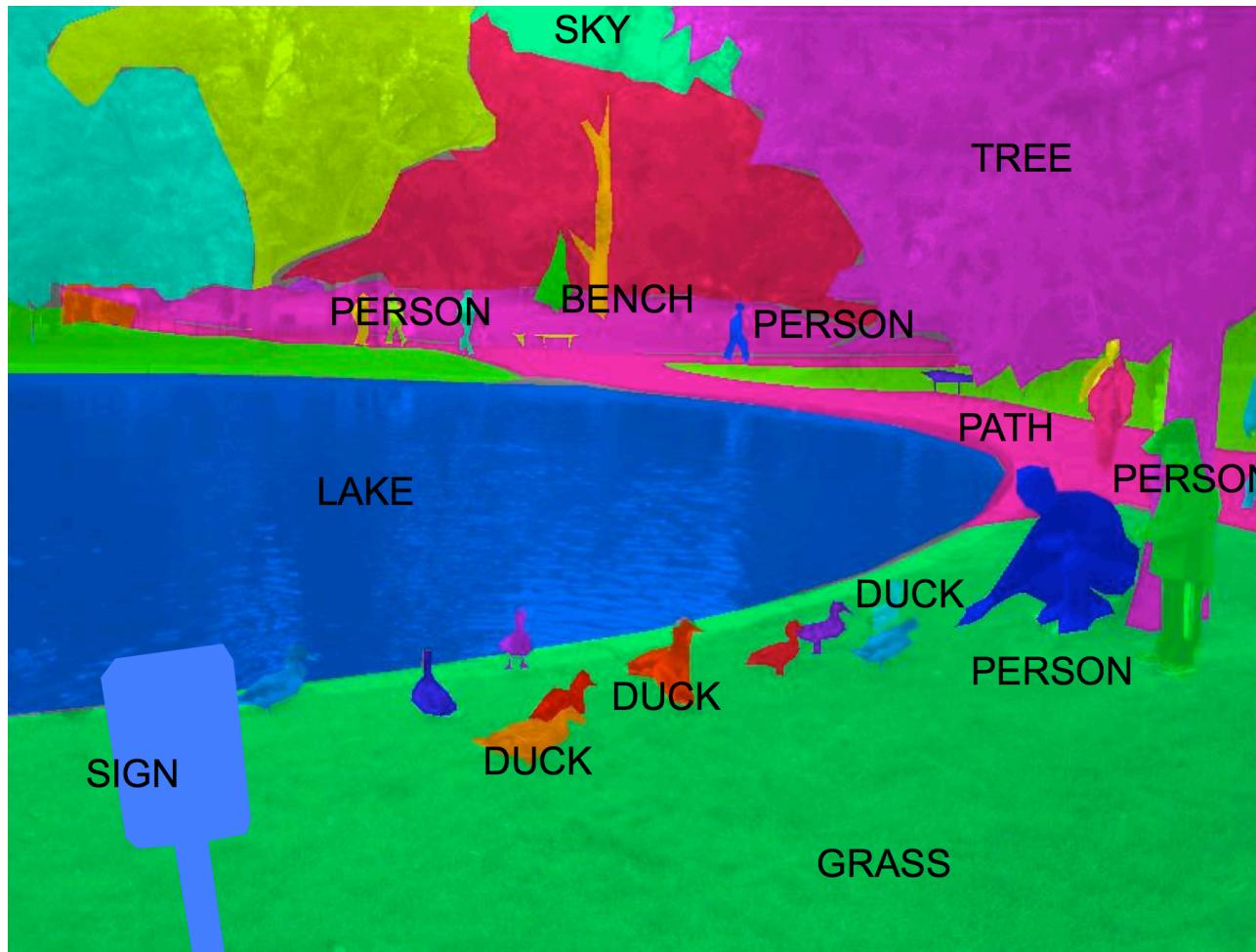
Semantic Segmentation: What we want



Segmentation



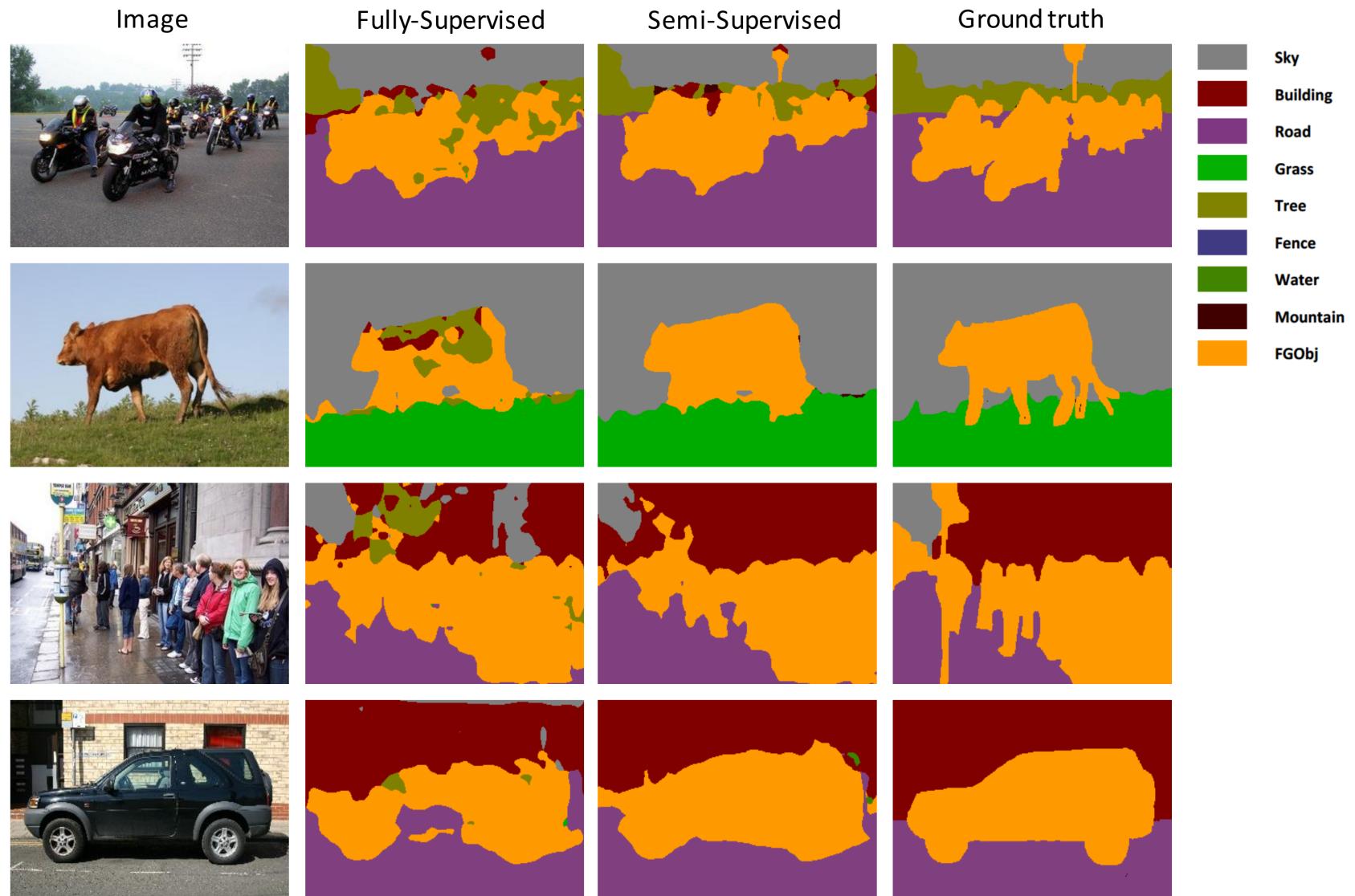
Segmentation



Segmentation



Semantic Segmentation: Results



*Semantic Segmentation Using GAN,
Nasim, Concetto, and Mubarak, 2017.*

Computer Vision Applications

Counting in Extremely Dense Crowd Images



Ground truth=634 Proposed Method by Idrees and Shah=640



Ground truth=1428

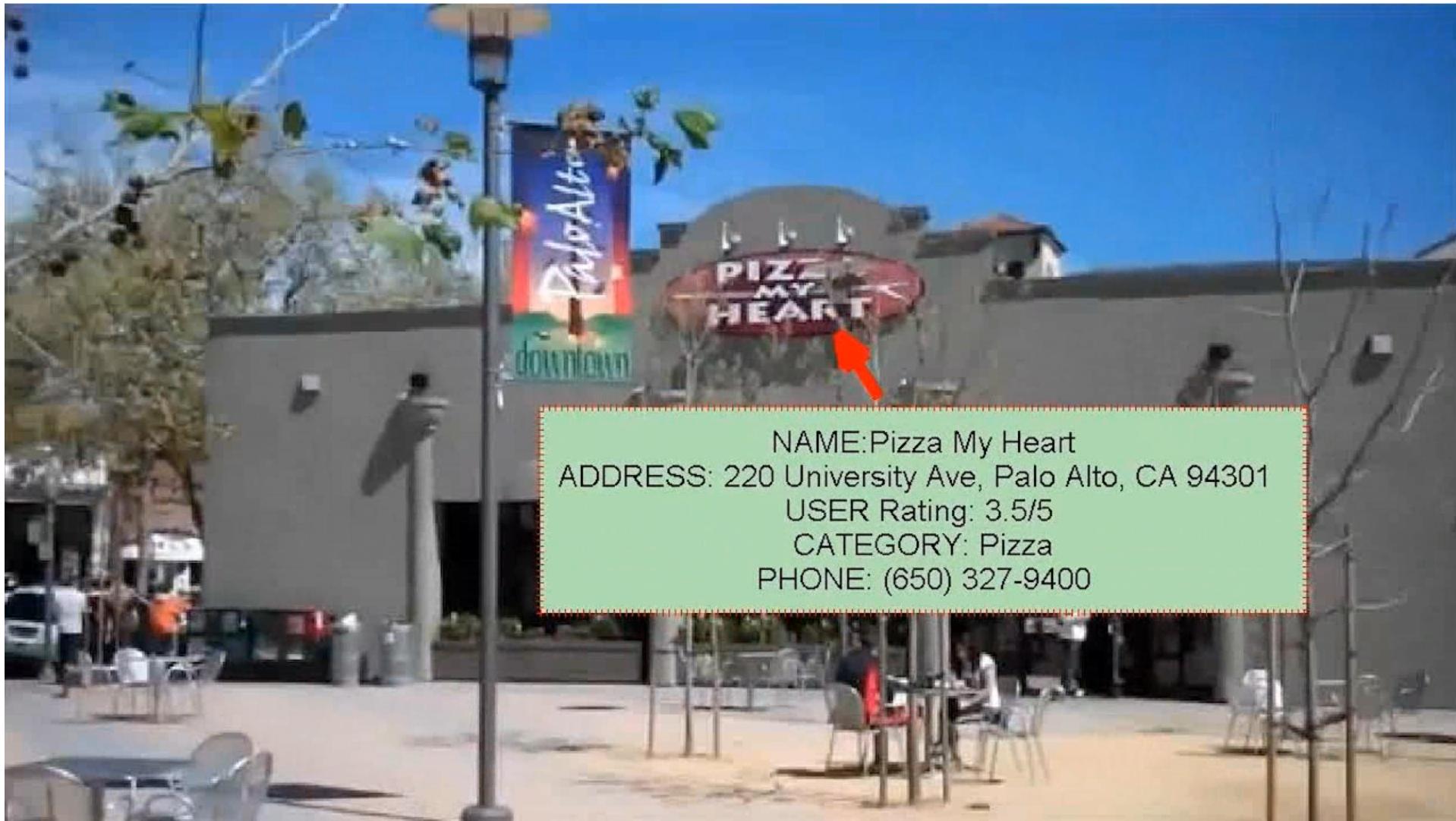
Proposed Method=1468



Ground truth=2319

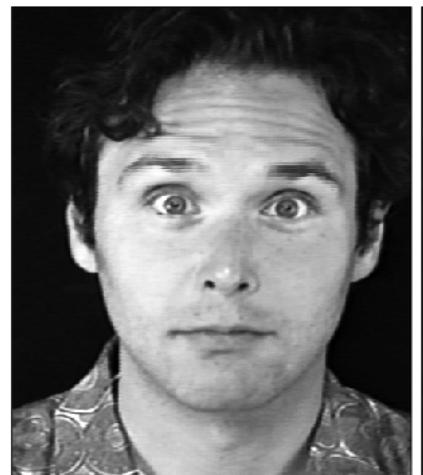
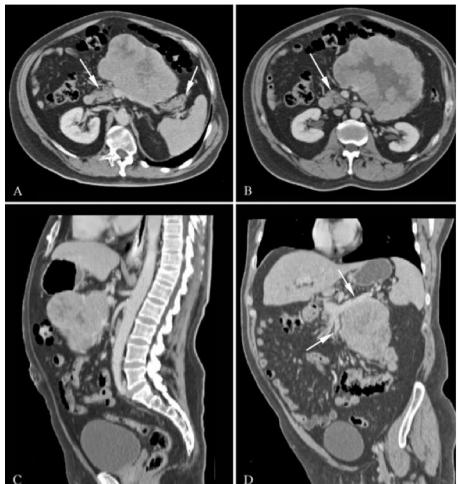
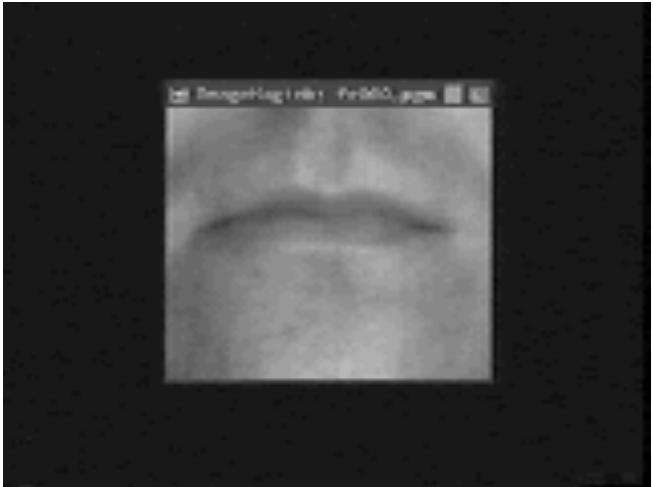
Proposed Method=2496

Visual Business Recognition



So many other applications...

- Face identification,
- Lip reading,
- Emotion detection,
- Medical Computer Vision / Analysis,



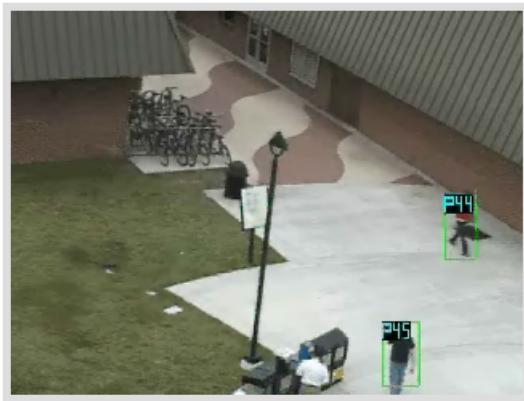
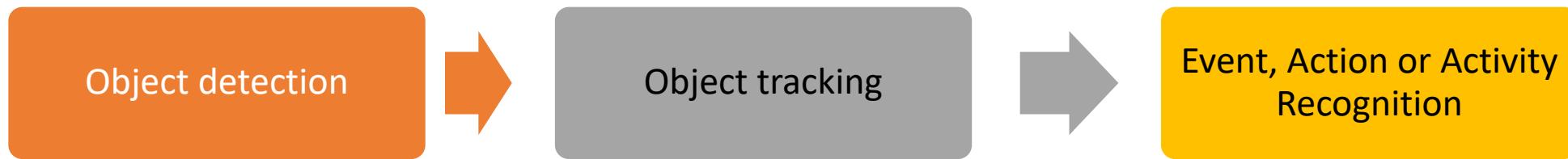
Surprised



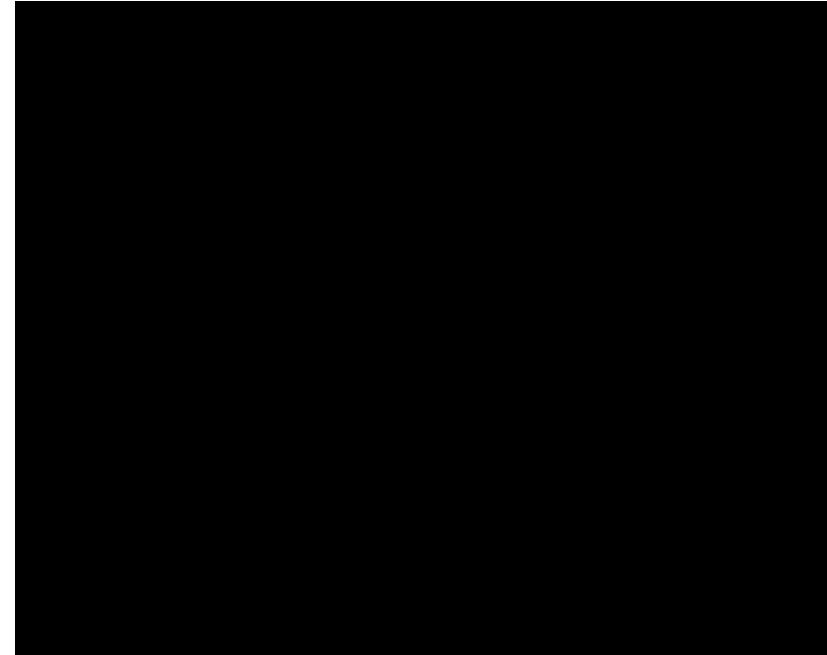
Happy

Video Analysis

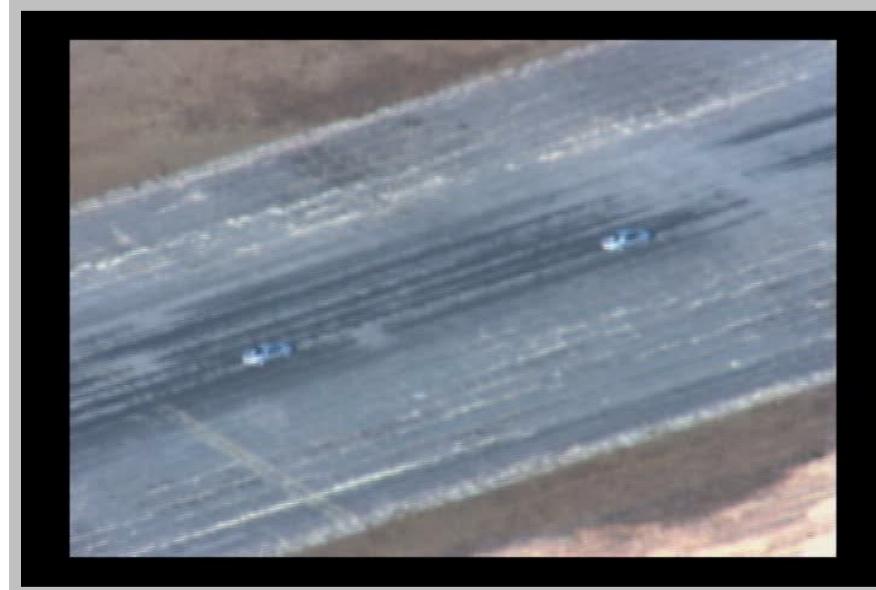
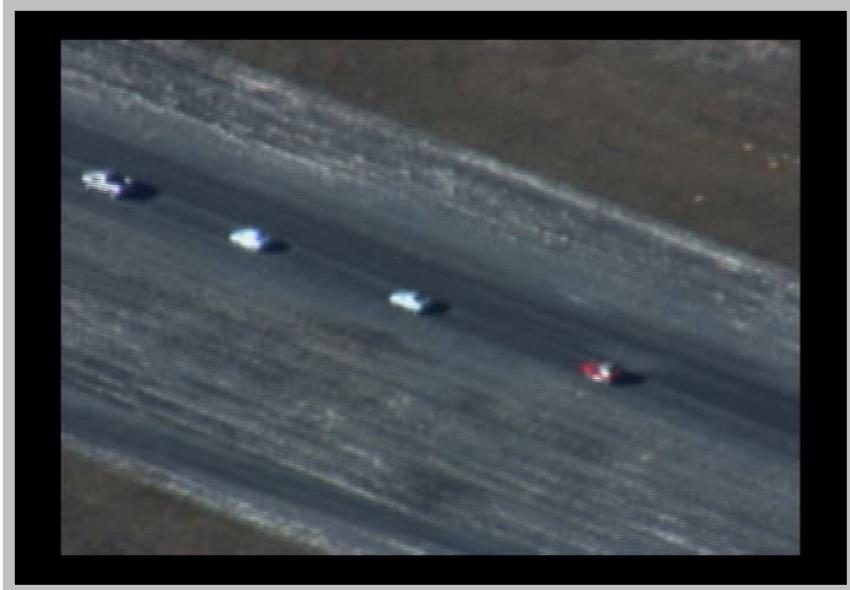
Video: image sequence (over time)



Object detection in videos



(Object) Tracking



- Deals with data association problem
 - (Correspondence problem)
- Correlates objects over time

Human Action Recognition

Action: A spatial and temporal pattern

9 actions, 142 videos (UCF database, Shah).



Bench Swing



Dive



Swing



Run



Kick



Lift



Ride

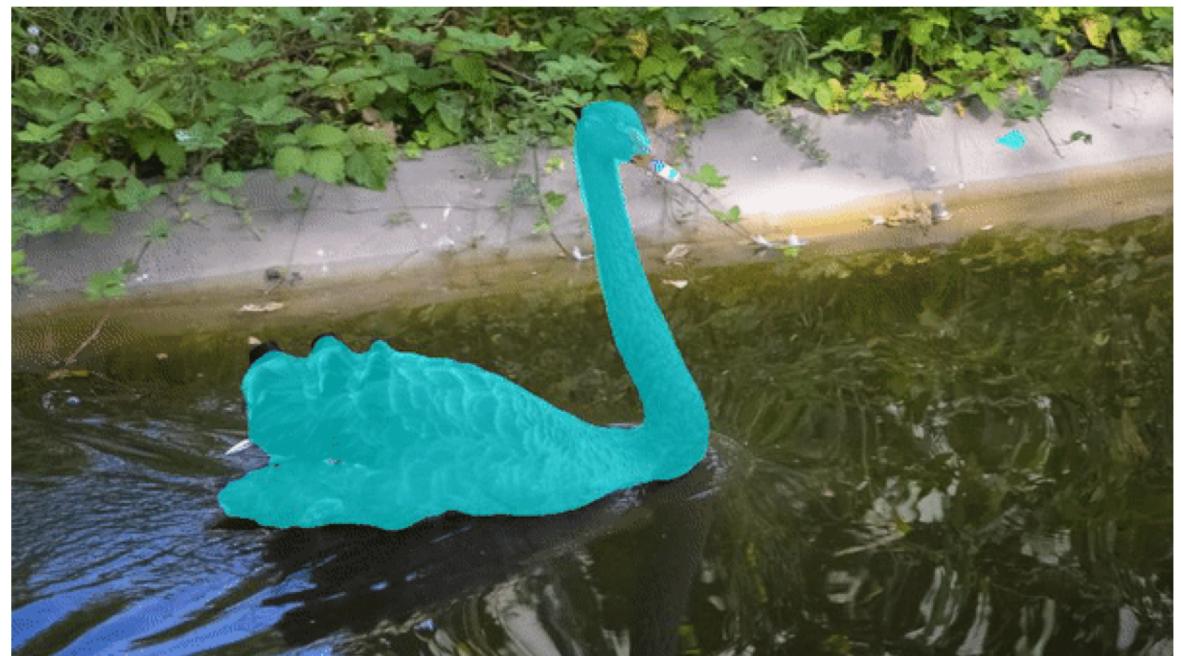


Golf Swing

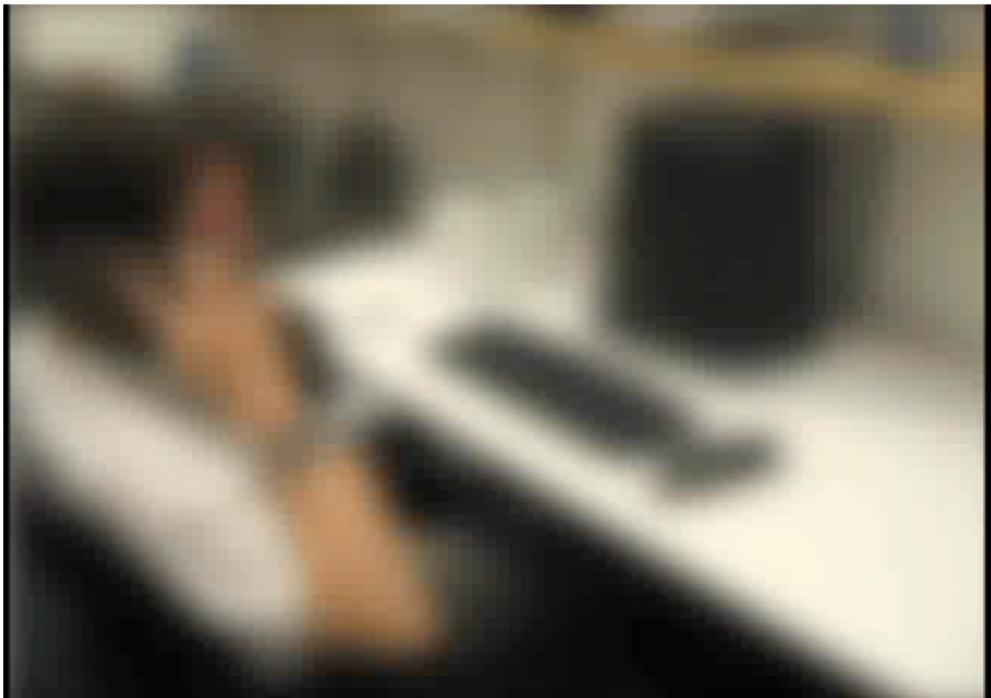


Skate

Actor – action segmentation



Test: what do you see in this image?



What do you see in this image?



Now can you see it?



Credit: Thompson, Basic Vision, Oxford Press, 2012.

Question:

How would you model (represent) an object in visual recognition? Ideas?

Remember the image:



```
01010100 01101000 01101001 01110011  
00100000 01101001 01110011 00100000  
01110100 01101000 01100101 00100000  
01110100 01110101 01110100 01101111  
01110010 01101001 01100001 01101100  
00100000 01110100 01101111 00100000  
01101100 01100101 01100001 01110010  
01101110 00100000 01100010 01101001  
01101110 01100001 01110010 01111001  
00101110 00100000 01001001 00100000  
01101000 01101111 01110000 01100101  
00100000 01111001 01101111 01110101  
00100000 01100101 01101110 01101010  
01101111 01111001 00100000 01101001  
01110100 00100001
```

This is what computers get (see?)!

This is what you see...

Our common units: Pixels

01010100 01101000 01101001 01110011
00100000 01101001 01110011 00100000
01110100 01101000 01100101 00100000
01110100 01110101 01110100 01101111
01110010 01101001 01100001 01101100
00100000 01110100 01101111 00100000
01101100 01100101 01100001 01110010
01101110 00100000 01100010 01101001
01101110 01100001 01110010 01111001
00101110 00100000 01001001 00100000
01101000 01101111 01110000 01100101
00100000 01111001 01101111 01110101
00100000 01100101 01101110 01101010
01101111 01111001 00100000 01101001
01110100 00100001

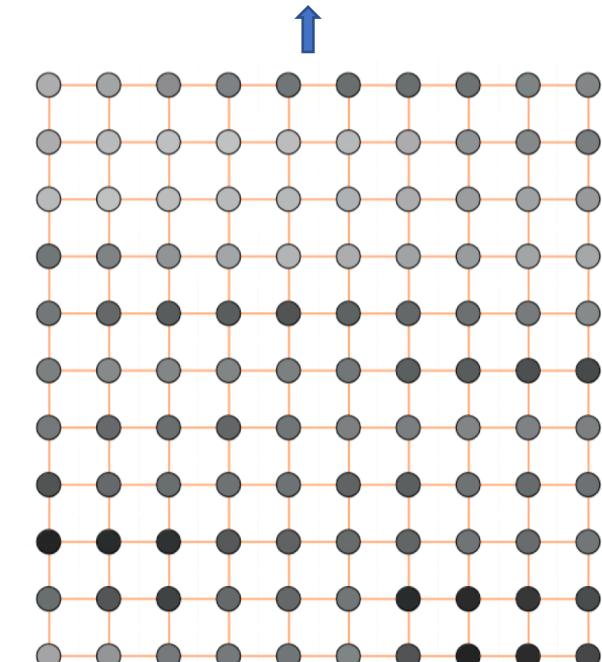
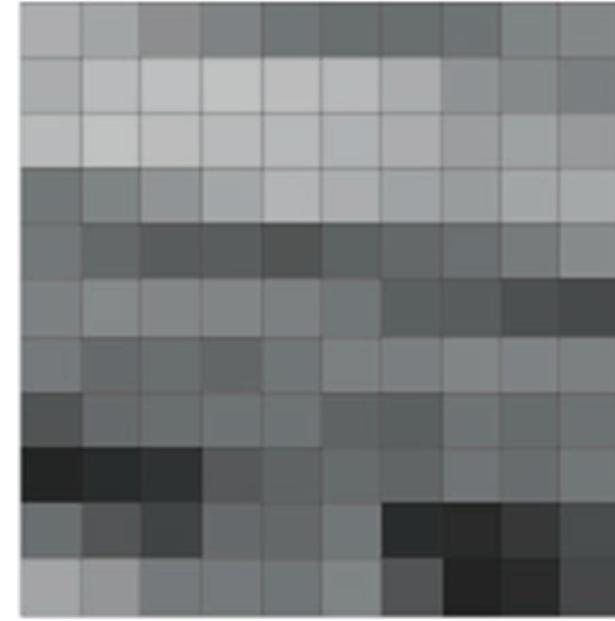
Binary

→

0	3	2	5	4	7	6	9	8
3	0	1	2	3	4	5	6	7
2	1	0	3	2	5	4	7	6
5	2	3	0	1	2	3	4	5
4	3	2	1	0	3	2	5	4
7	4	5	2	3	0	1	2	3
6	5	4	3	2	1	0	3	2
9	6	7	4	5	2	3	0	1
8	7	6	5	4	3	2	1	0

→

Decimal - grid



Grayscale image

Image Types: (Gray)Scalar and Binary

- A scalar image has integer values

$$u \in \{0, 1, \dots, 2^a - 1\}$$

a: level (bit)



Ex. If 8 bit (a=8), image spans from 0 to 255

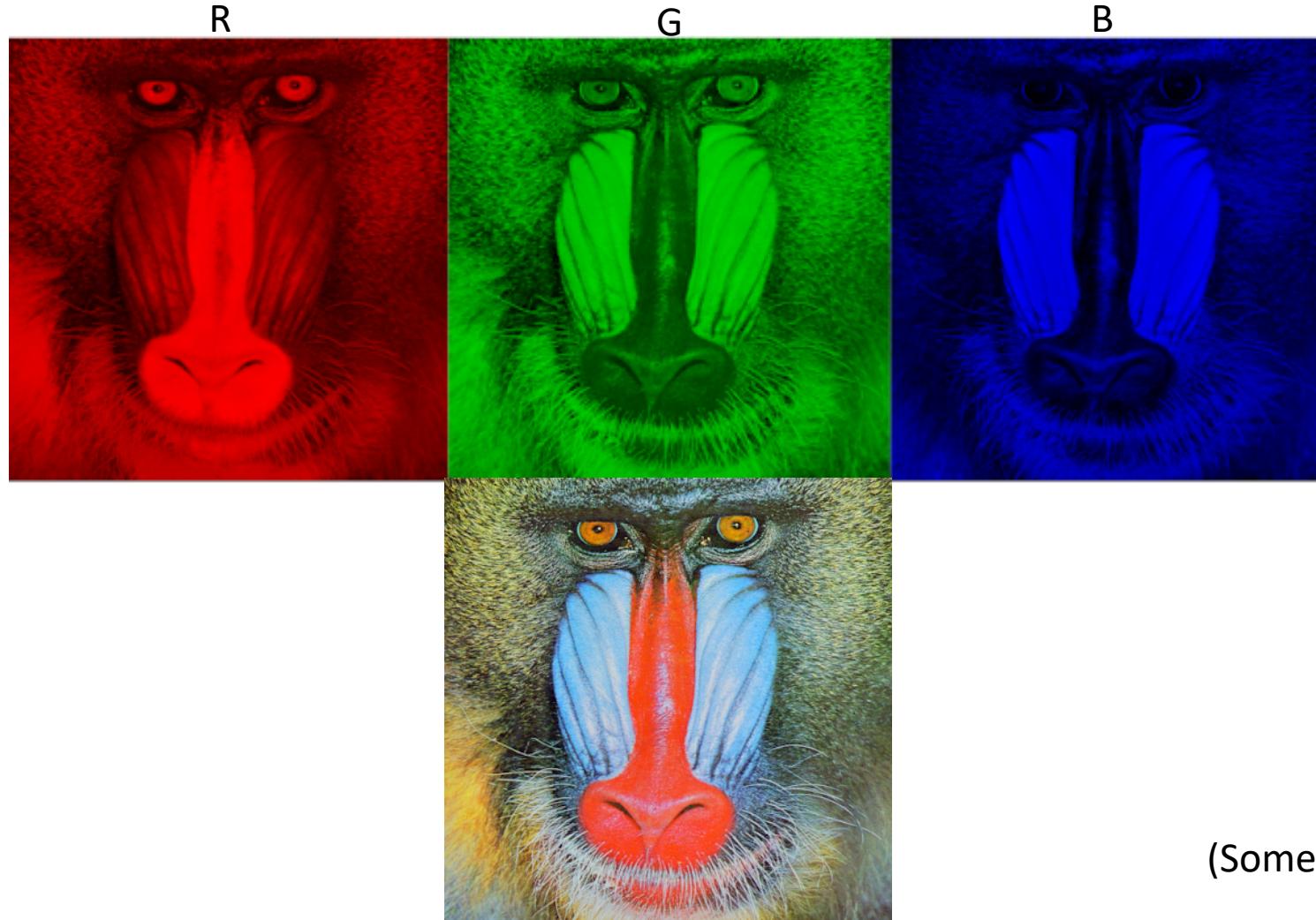
0 black

255 white

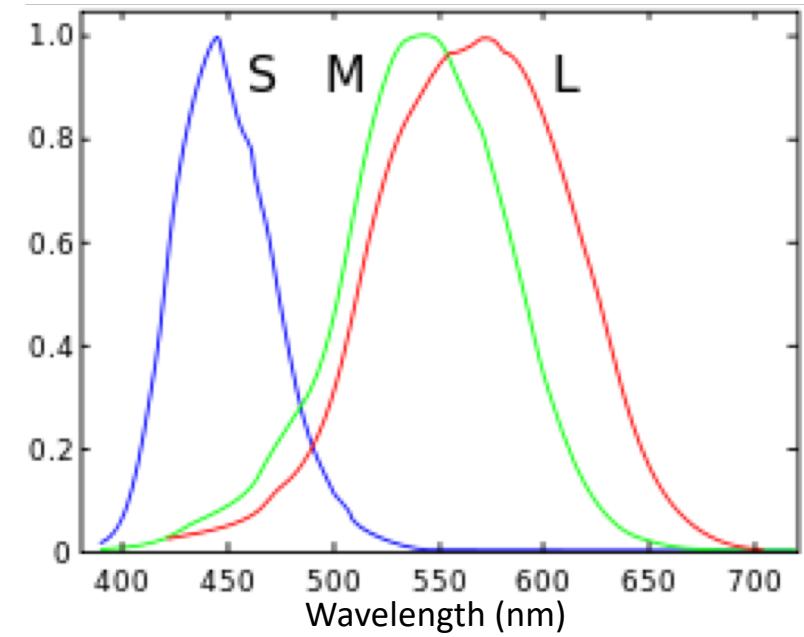
Ex. If 1 bit (a=1), it is binary image, 0 and 1 only.

Image Type: RGB (red, green, blue)

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



Human Cone-cells (normalized)
responsivity spectra

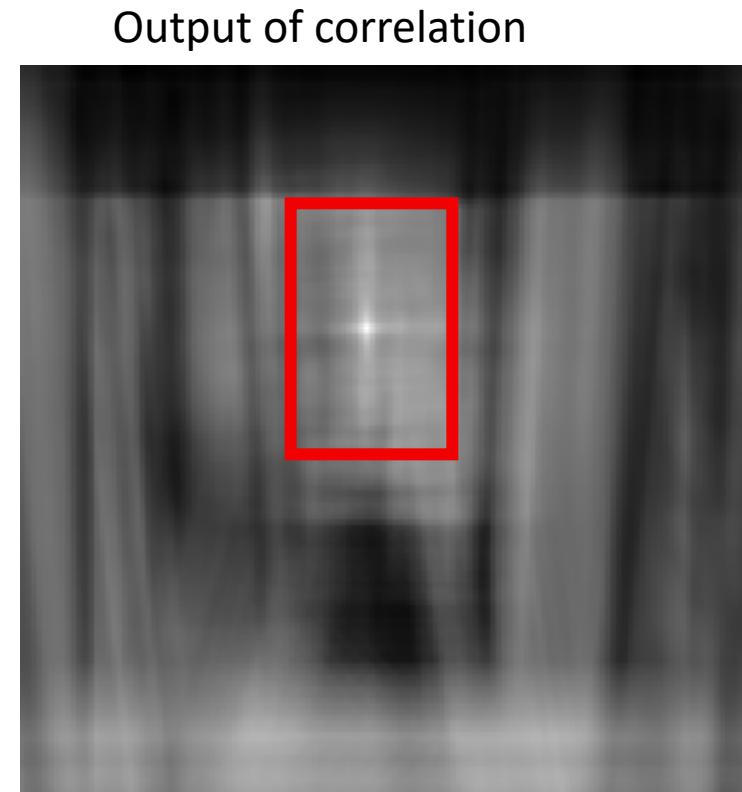


(Some people might have 4 cone-types!)

So.... How do we detect an object in an image?

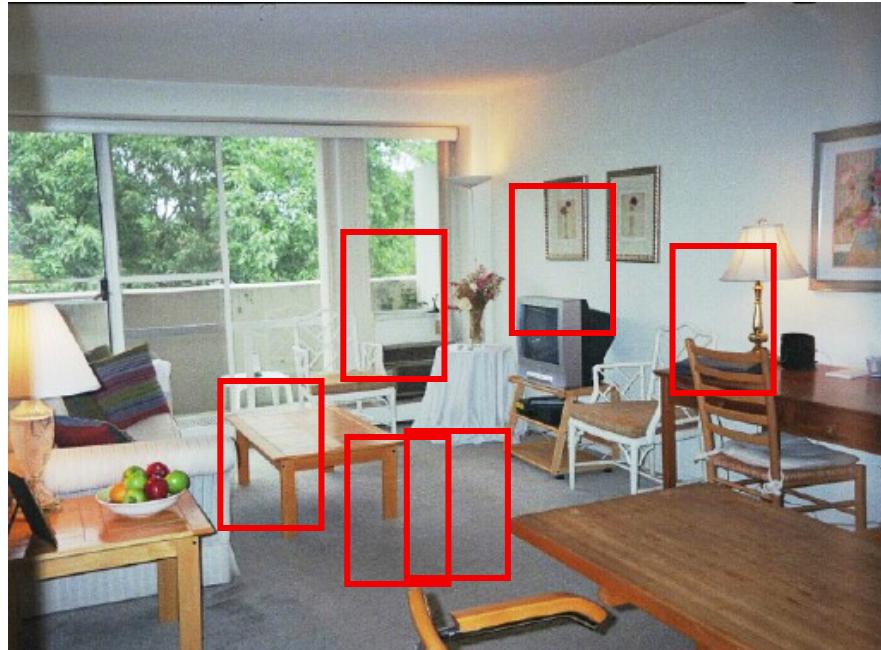
Naïve approach: Template Matching

This is a chair



Template Matching

Find the chair in this image



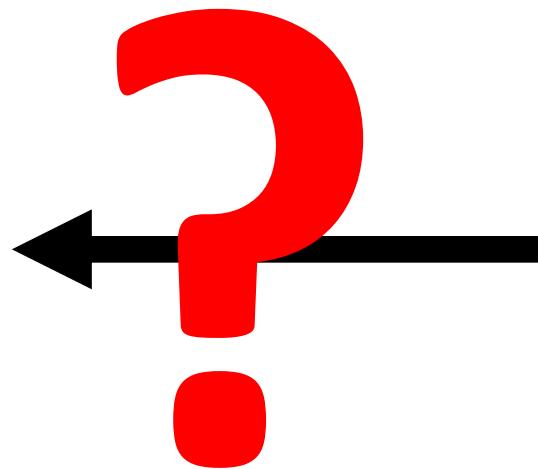
Epic fail!

Simple template matching is not going to make it

Idea:

- Instead of comparing raw image pixels:
 - first map those pixels into another (more robust) form,
 - and then compare those mapped forms.
 - Finally, select the closest image map (how do you define “closest”? Metrics).
- Features
 - Examples: compute edges, compute color histograms, Gradients, HOG, SIFT, ...

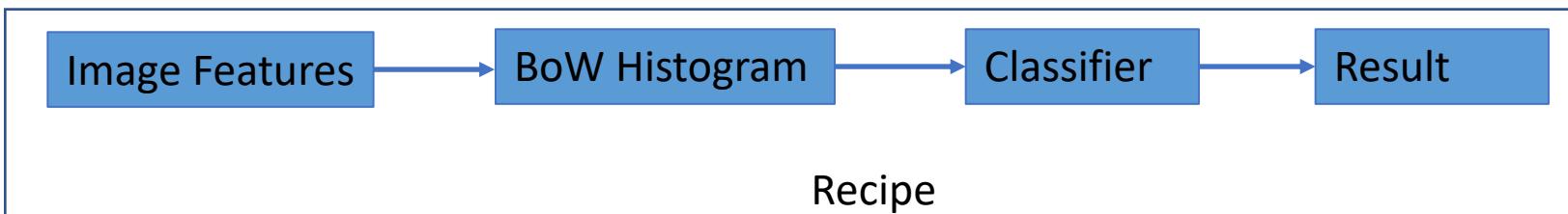
“Bag-of-Words” Representation



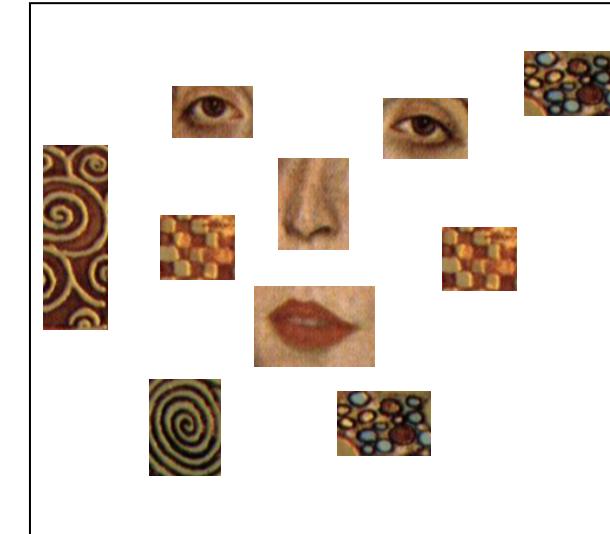
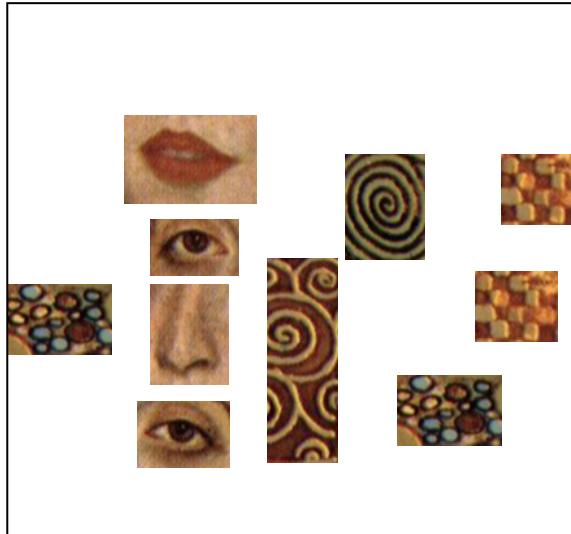
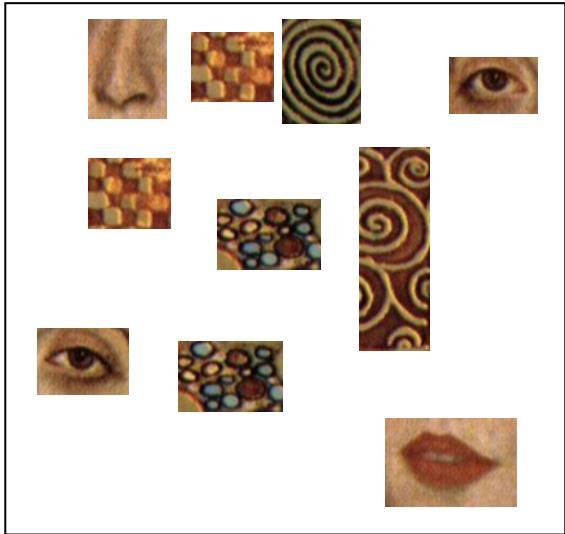
“Bag-of-Words” (BoW) Histograms



Image



BoW Representation



- All have equal probability for bag-of-words methods,
- Location (spatial) information is important but lost.

A Brief History

The early optimism (1960-1970)

MASSACHUSETTS INSTITUTE OF TECHNOLOGY

PROJECT MAC

Artificial Intelligence Group
Vision Memo. No. 100.

July 7, 1966

THE SUMMER VISION PROJECT

Seymour Papert

The summer vision project is an attempt to use our summer workers effectively in the construction of a significant part of a visual system. The particular task was chosen partly because it can be segmented into sub-problems which will allow individuals to work independently and yet participate in the construction of a system complex enough to be a real landmark in the development of "pattern recognition".

50 Years Ago

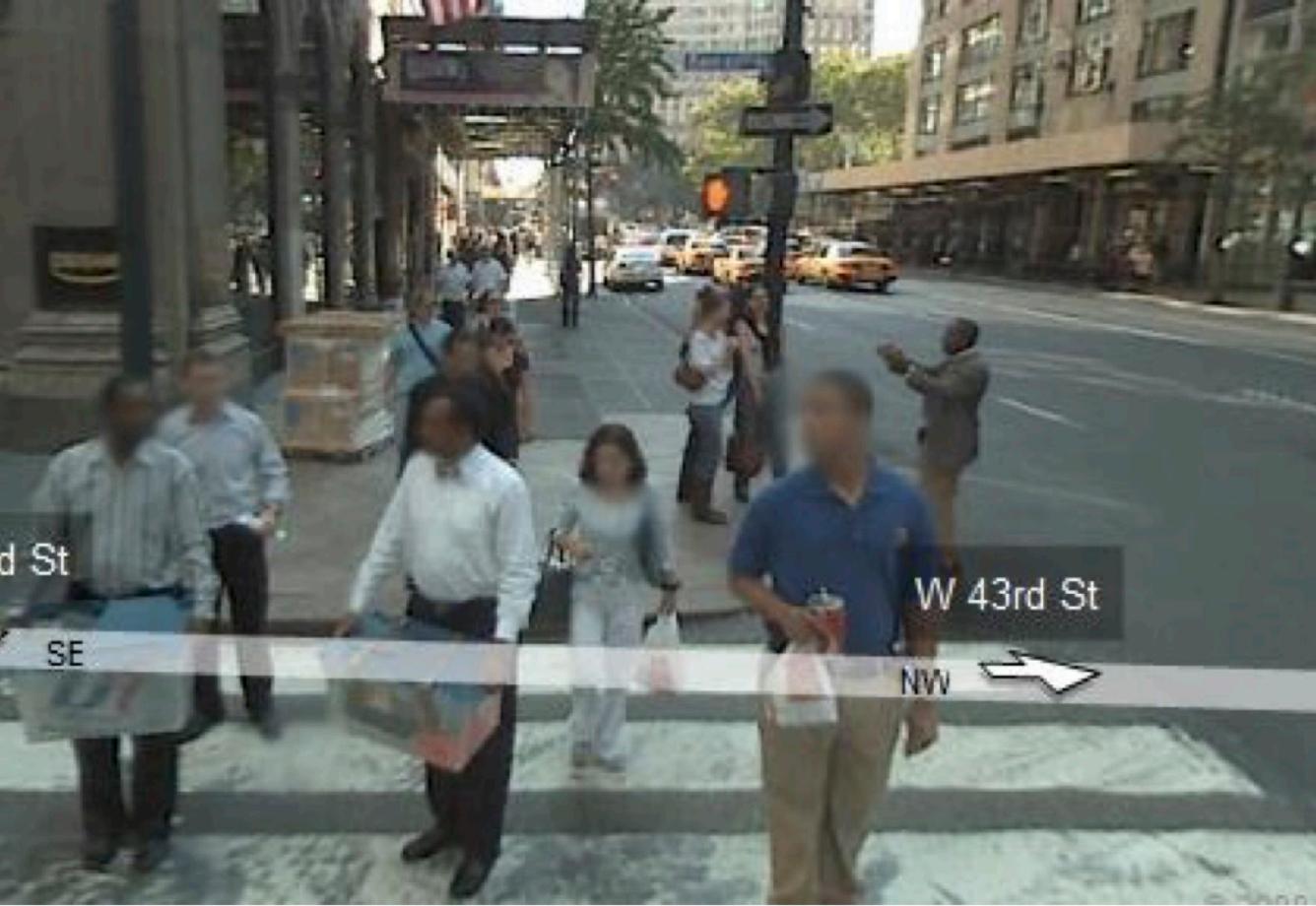


25 Years ago

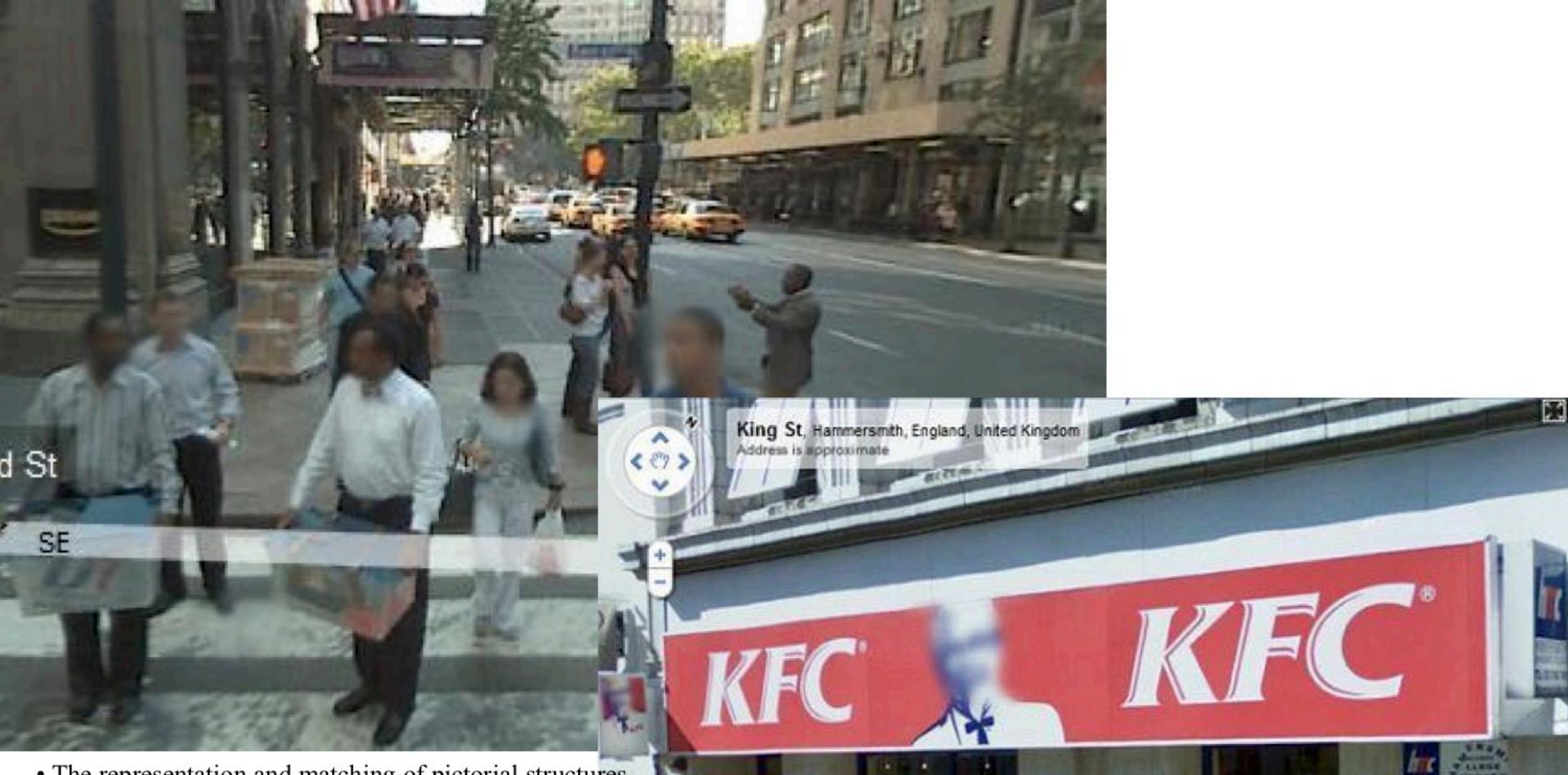


15 Years ago



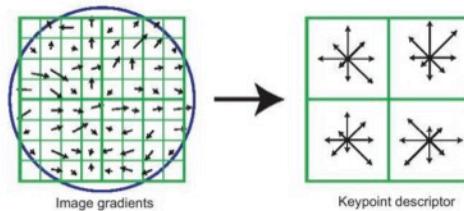


- The representation and matching of pictorial structures
Fischler, Elschlager (1973).
- Face recognition using eigenfaces M. Turk and A.
Pentland (1991).
- Human Face Detection in Visual Scenes - Rowley, Baluja,
Kanade (1995)
- Graded Learning for Object Detection - Fleuret, Geman
(1999)
- Robust Real-time Object Detection - Viola, Jones (2001)
- Feature Reduction and Hierarchy of Classifiers for Fast
Object Detection in Video Images - Heisele, Serre,
Mukherjee, Poggio (2001)
-

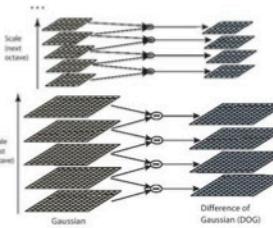


- The representation and matching of pictorial structures
Fischler, Elschlager (1973).
- Face recognition using eigenfaces M. Turk and A. Pentland (1991).
- Human Face Detection in Visual Scenes - Rowley, Baluja, Kanade (1995)
- Graded Learning for Object Detection - Fleuret, Geman (1999)
- Robust Real-time Object Detection - Viola, Jones (2001)
- Feature Reduction and Hierarchy of Classifiers for Fast Object Detection in Video Images - Heisele, Serre, Mukherjee, Poggio (2001)
-

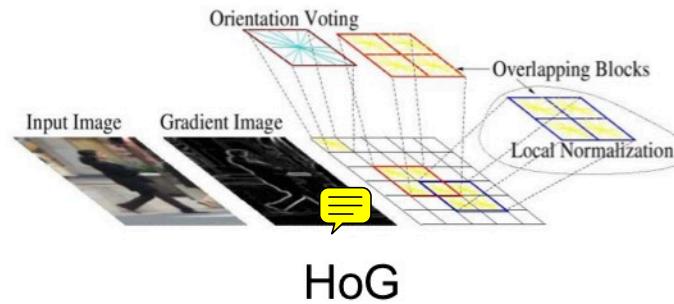
Advances in computer vision



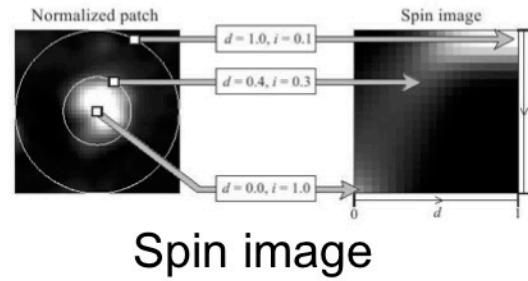
SIFT



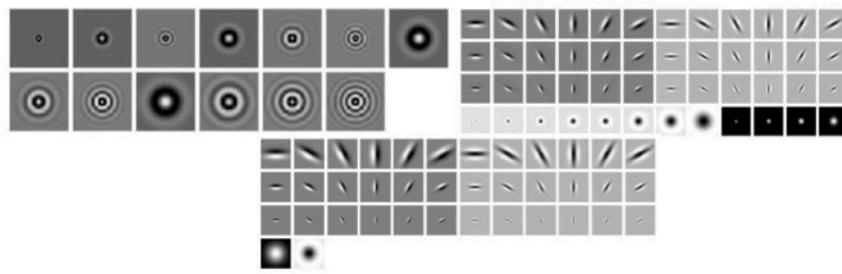
GIST



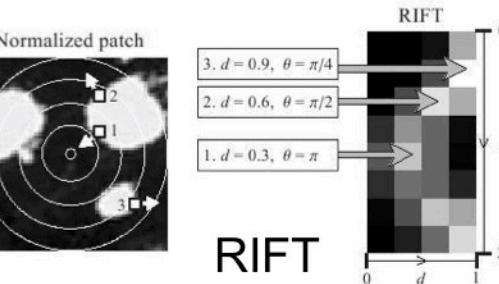
HoG



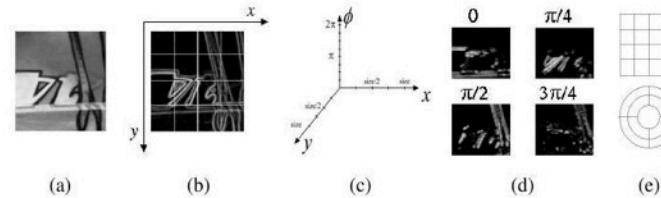
Spin image



Textons

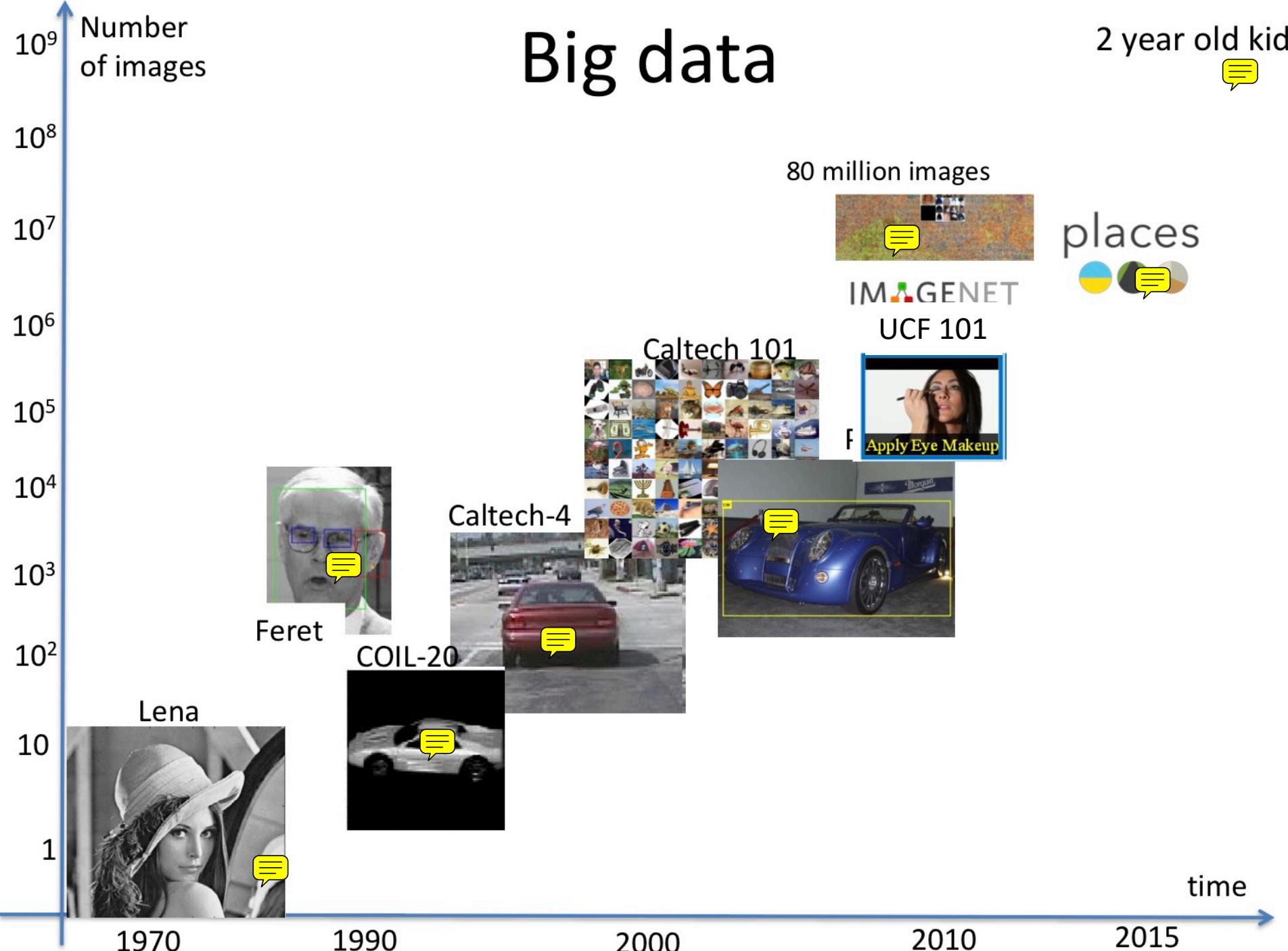


RIFT



GLOH

Big data



The time of big data



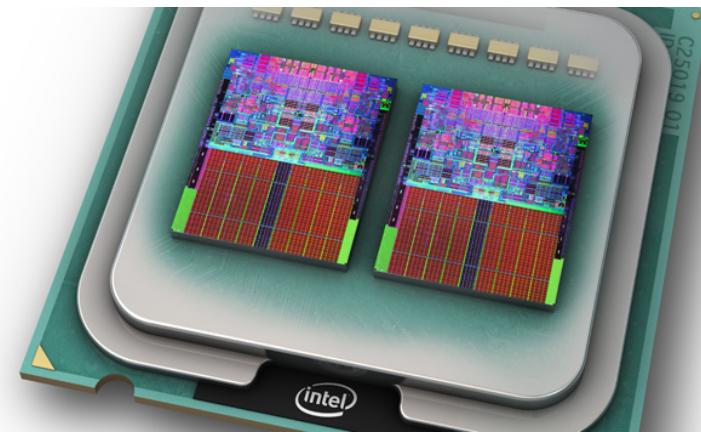
Distributed / GPU / Cloud Computing



Google Cloud



Microsoft Azure



QUESTIONS?

This lecture contains content from:

- Antonio Torralba (MIT),
- Ulas Bagci (UCF),