

Fall 2022

INTRODUCTION TO COMPUTER VISION

Atlas Wang

Assistant Professor, The University of Texas at Austin

Overview of Course Logistics

- We meet on Tuesday & Thursday 5:00-6:30pm (ETC 2.136)
 - Class format: in-person (sorry, no video streaming/record)
 - Do I have to come to the classroom?
 - Can I audit?
 - After-class communication: Slack (link sent) IMPORTANT!



- Class materials are distributed on Course Webpage (NOT Canvas): https://vita-group.github.io/fall_22.html
- We do not follow any textbook closely. Instead we will have many "recommended materials".

Overview of Course Logistics

- Instructor Office Hour: Wednesday 10:00am 11:00am, meet at Office EER 6.886
- This class has two TAs:
 - Dejia Xu, dejia@utexas.edu, Office Hour: Monday 4:00-5:00pm
 - Wenyan Cong, wycong@utexas.edu, Office Hour: Friday 4:00-5:00pm
 - Both TA slots: meet outside EER O's Campus Café, outdoor seating area
- Which office hour should I come to?
- Online Q&A: anytime, just ask on Slack!

Grading

- Homework: 20%
 - There will be 4 mandatory written or machine assignments, 5% each.
 - One 5th homework (5%) will be given as "optional" bonus
 - HOMEWORK 0 out today! Due next Monday (8/29) EOD
- Mid-term exam: 30% (time/format TBD)
- Final Project: 50%
 - Proposal (10%) <u>Due by the end of Week 8 (10/16 Sunday</u>): 2-Page report, including project title, team member, problem description, preliminary literature survey, the proposed technical plan, and references
 - Presentation (10%): Be prepared to be challenged by your peers and the instructor
 - Time TBD, usually in the last week of semester, before the final exam week
 - Code review (10%): Write clean, well-documented and runnable codes, PLEASE
 - Final Report (20%): (8+1)-page report following the standard CVPR paper template (and quality level)
 - Template file: http://cvpr2020AuthorKit.zip

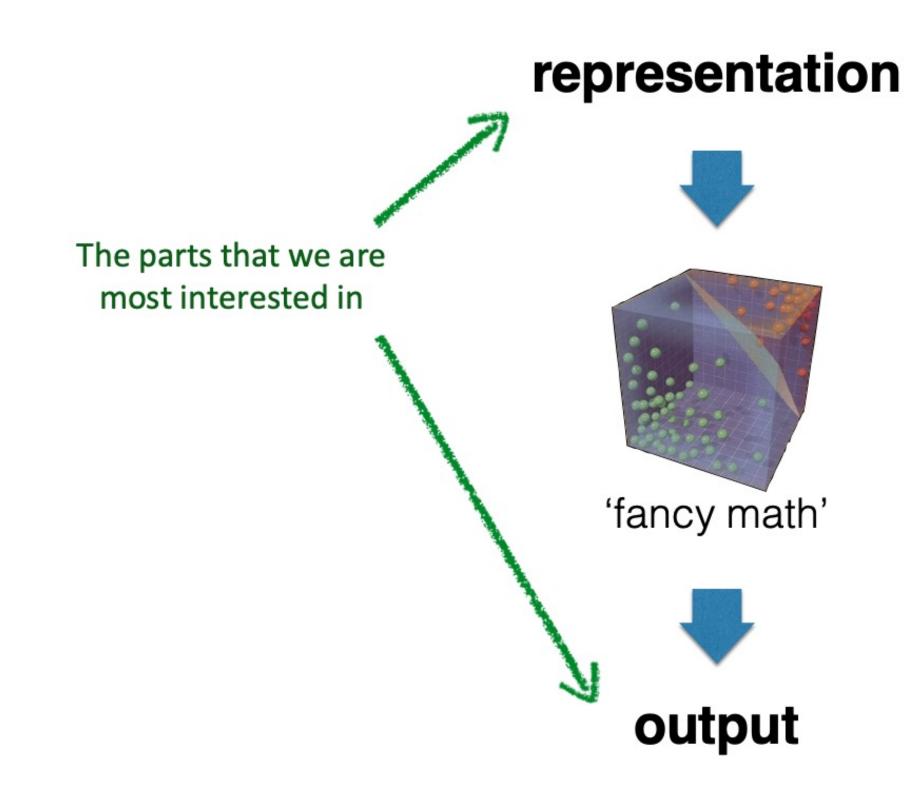
What is Computer Vision?

- An interdisciplinary field that deals with how computers can be made for gaining holistic understanding from digital images or videos.
- From the engineering perspective, it seeks to automate tasks that the human visual system can do.

Computer Vision as Input-Output System:

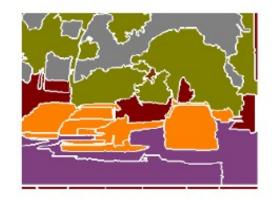
- Input: images or video
- Output (ideally): description or understanding of the visual world, in a "human" way
- Outputs (practically): reconstructing, measuring, classifying, interpreting...

A Conceptual Visual Perception Pipeline

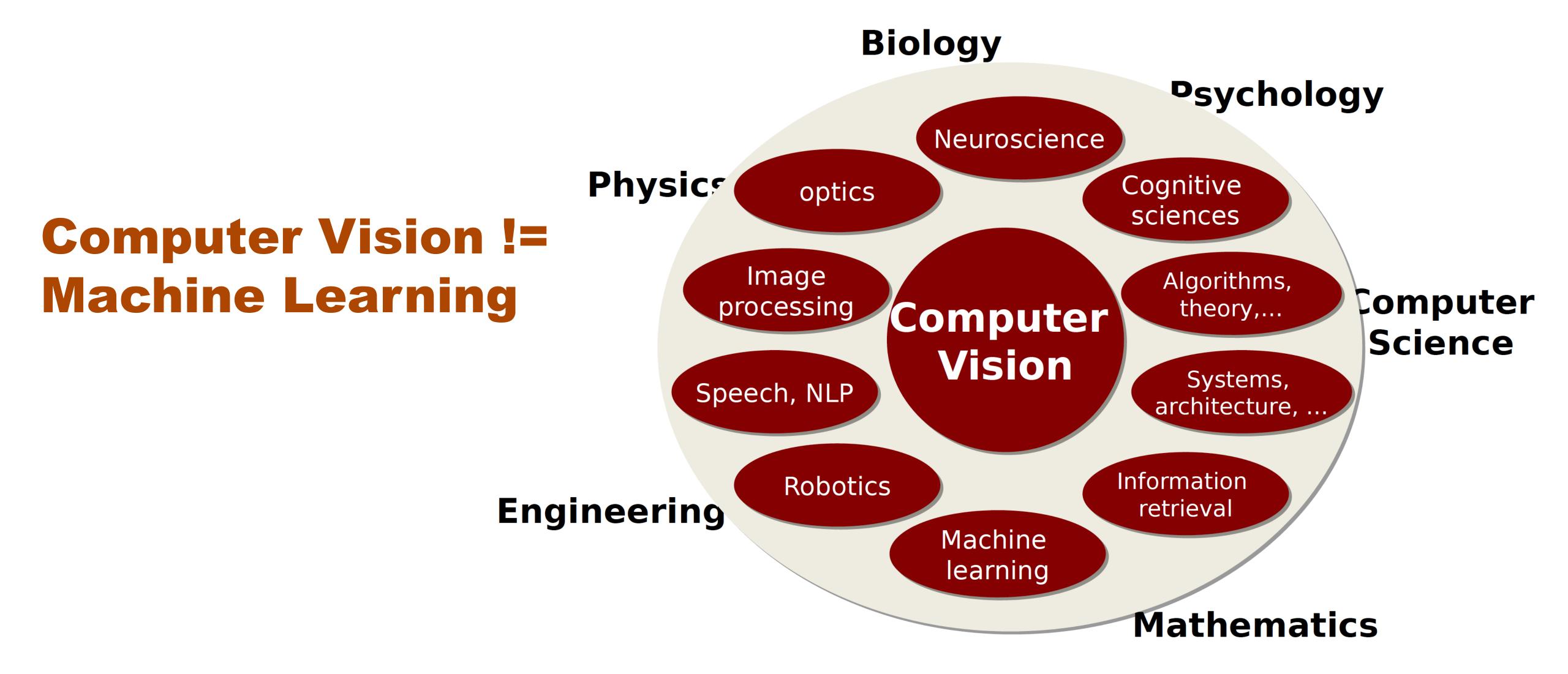




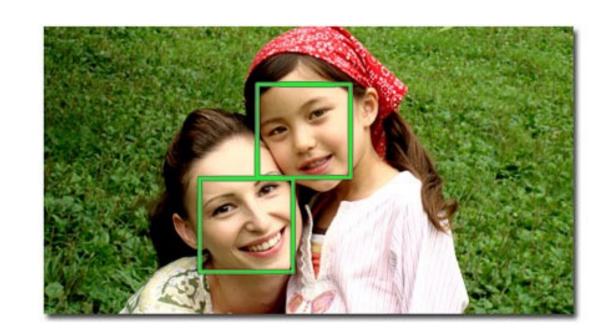
what should we look at? (image features)



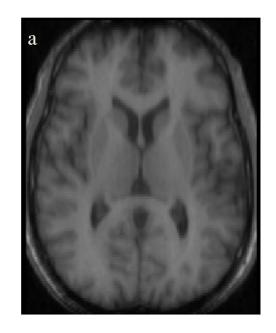
what can we understand? (semantic segmentation)



Computer Vision has SO MANY applications

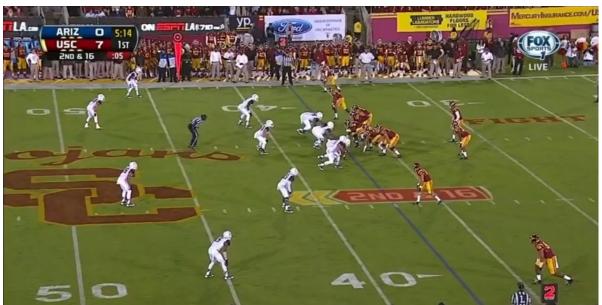


Face Detection/Smile recognition



Medical Image Understanding

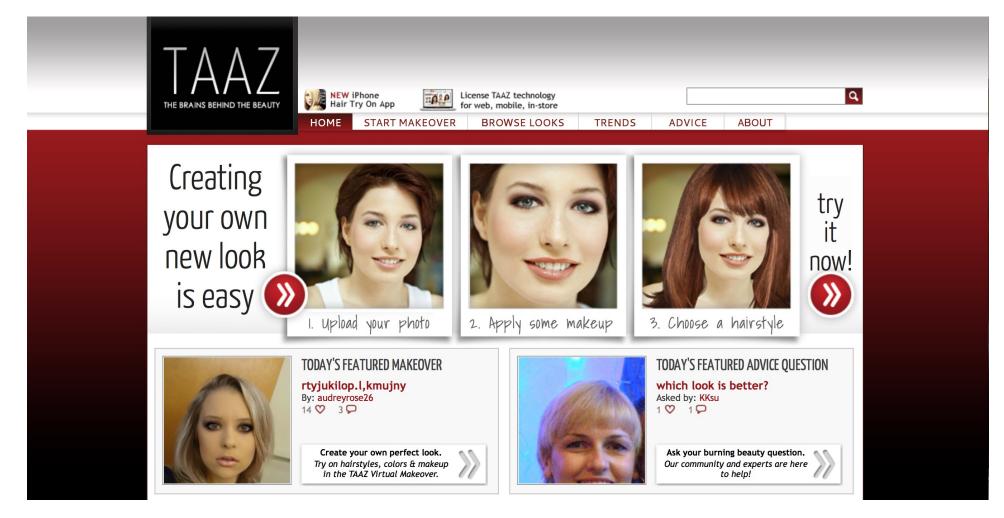




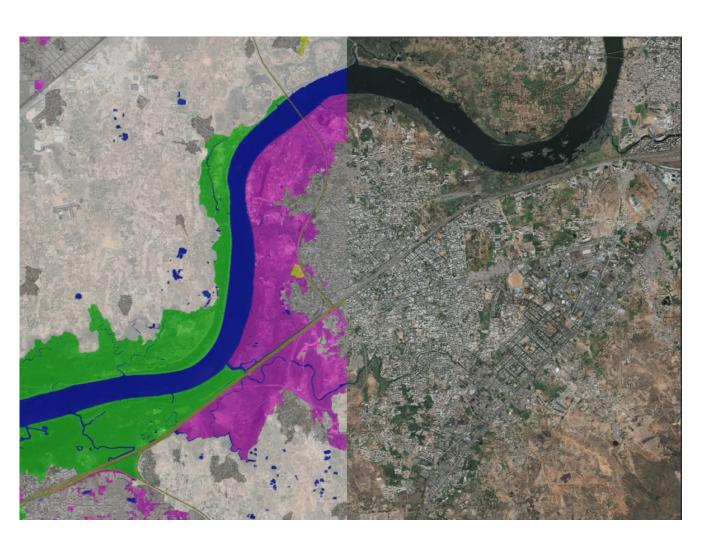
Tracking in Sports



Self-driving cars



Face Makeover/Virtual try-on



Remote sensing/earth mapping



Pose estimation (esp. fall detection)

... and Even More Open Challenges



Amazon's camera-equipped Echo Look raises new questions about smart home privacy

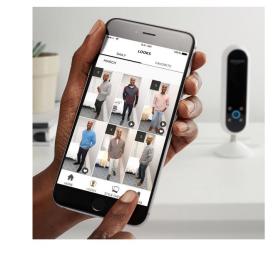


Startups Apps

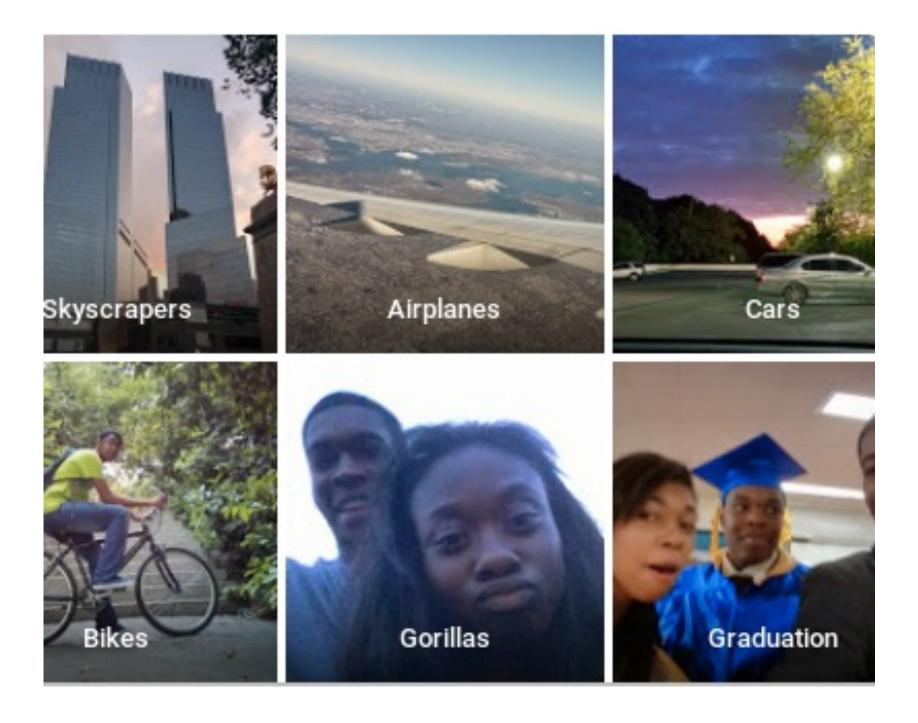
Apps Gadgets

Brian Heater @bheater / Apr 26, 2017









MASSACHUSETTS INSTITUTE OF TECHNOLOGY PROJECT MAC

TEXAS ELECTRICAL AND COMPUTE

Artificial Intelligence Group Vision Memo. No. 100. July 7, 1966

Do you know?

The first "Computer Vision" work in this world was originally a summer project given to an MIT undergraduate student

THE SUMMER VISION PROJECT

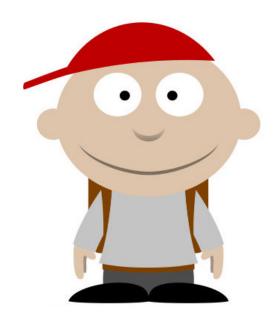
Seymour Papert

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".

Computer Vision Research History: My (probably approximately correct) summary

- Late 1960s: CV was born = a branch of human vision and cognition research (bio-inspired CV)
- **1970s**: CV = estimate 3D structures from 2D images (*physically-grounded CV*)
- 1980s: more rigorous math concepts such as scale space, texture analysis, contour models, as well as the emergence of optimization and inference methods
- **Early-to-mid 1990s:** camera calibration, multi-view stereo, scene reconstruction, image segmentation, the big boom of statistical learning methods
- Late 1990s: bridging CV and graphics: rendering, morphing, stitching...
- 2000s and after: ML (graphical models, sparsity & low-rank), and finally Deep Learning ...

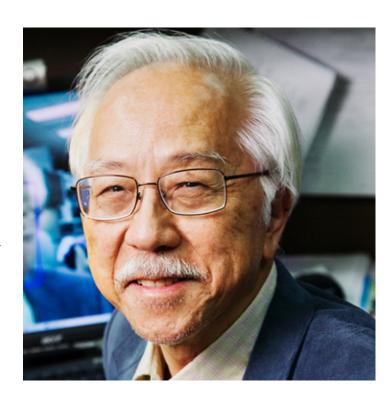
After 56 Years...Computer Vision is Still Tough!



Atlas Wang

Hey Tom, What do you see as the biggest problem in computer vision?

One biggest problem of computer vision is — human never see in pixels!



Prof. Thomas S. Huang (1936 - 2020), ECE@UIUC

"A founding father in computer vision"



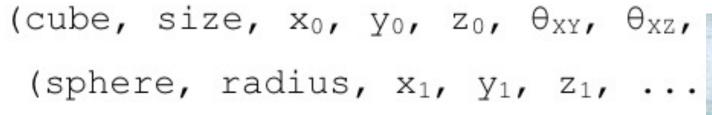
When David Marr at MIT moved into computer vision, he generated a lot of excitement, but he hit up against the problem of knowledge representation; he had no good representations for knowledge in his vision systems.

— Marvin Minsky —

AZ QUOTES

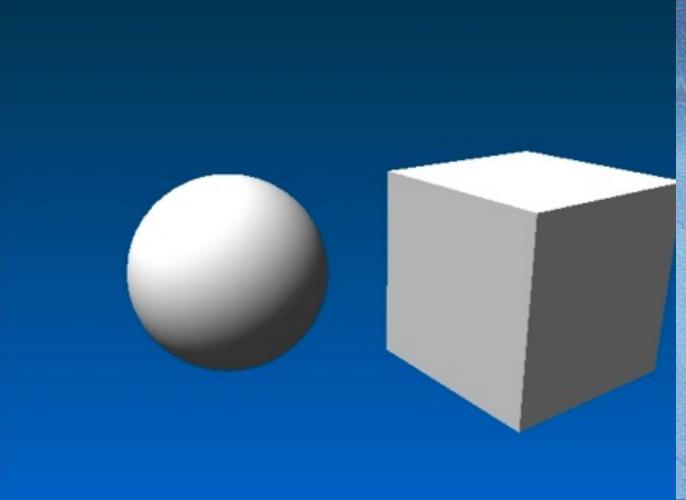
- Situation much the same as Al:
 - Some fundamental algorithms
 - Large collection of hacks / heuristics
- CV research is hard and "never ending"
 - Especially at high level, physiology unknown
 - Requires integrating many different methods
 - Requires reasoning and understanding: "Al completeness"

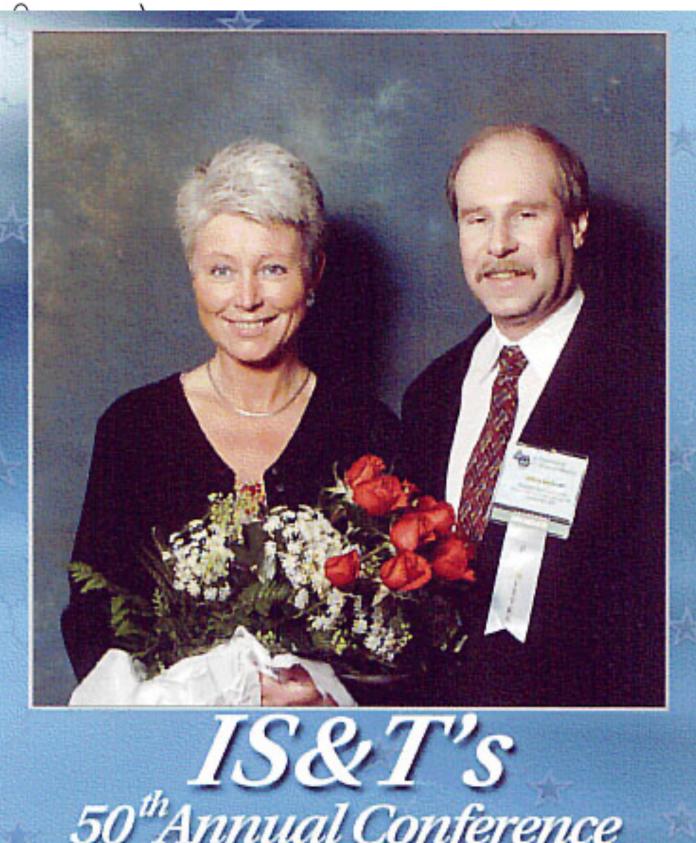
TEXAS ELECTRICAL AND COMPUTER ENGINEERING



Computer Graphics

Comput Vision





IS&T'S 50thAnnual Conference



Computer Vision and Image Processing are significantly overlapped in their tools (http://www.lenna.org/)

Computer Vision and Computer Graphics are often viewed as "inverse operations"

Marr's Tri-Level Hypothesis for Vision

David Marr integrated results from psychology, artificial intelligence, and neurophysiology into new models of visual processing, creating the field of Computer Vision.

- Computational level: what does the system do (e.g.: what problems does it solve or overcome) and similarly, why does it do these things -- What is the problem?
- Algorithmic level (a.k.a. representational level): how does the system do what it does, specifically, what representations does it use and what processes does it employ to build and manipulate the representations -- How to solve the problem?
- Implementational level (a.k.a. physics level): how is the system physically realized (in the case of biological vision, what neural structures and neuronal activities implement the visual system) -- How the above are done in a computer or a brain?







David Marr

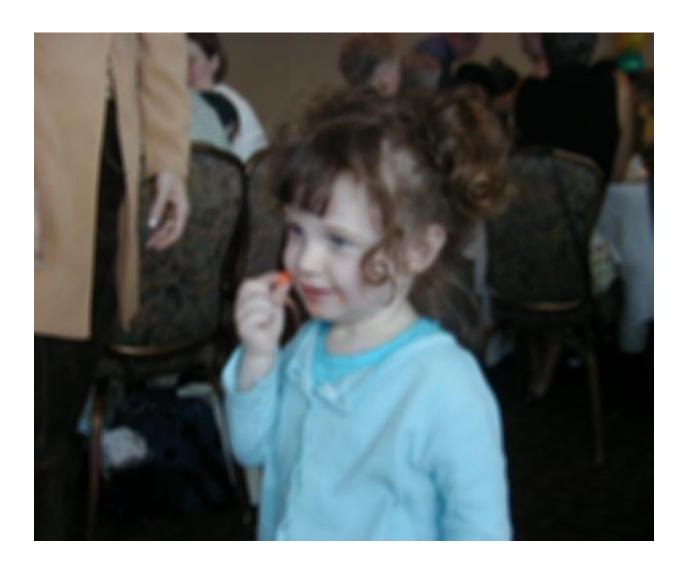
Shimon Ullman

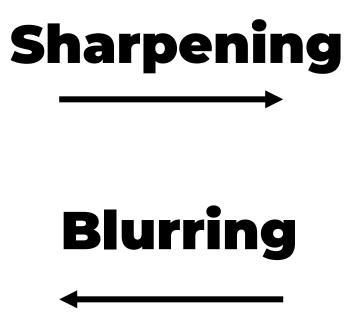
AFTERWORD BY

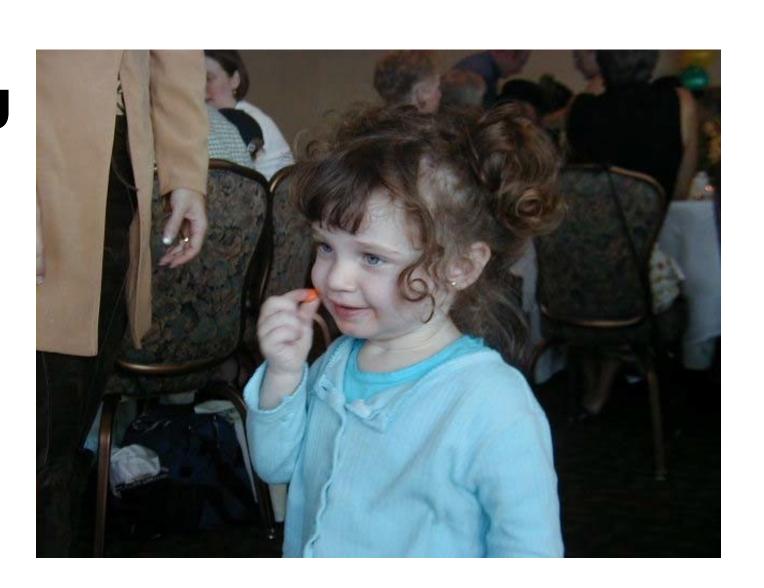
Tomaso Poggio

Three Stages in Computer Vision

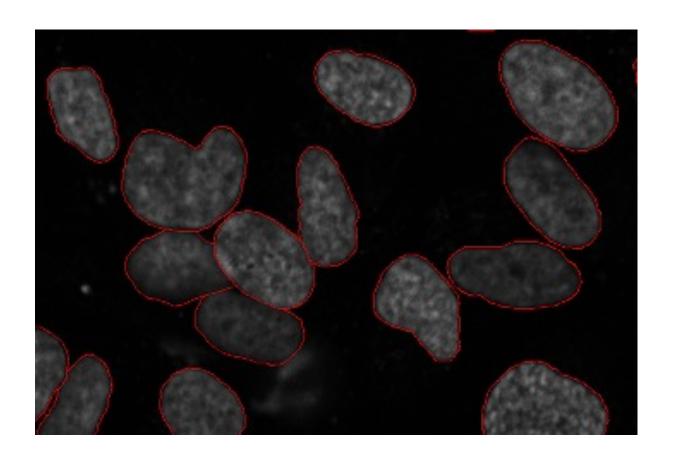
- Low-Level: Image to image (enhancement, edge detection...)
 - Largely overlapped with signal or image "reconstruction" & "filtering"
 - Directly interface with image formulation, often considered as "pre-processing" for CV tasks







Three Stages in Computer Vision

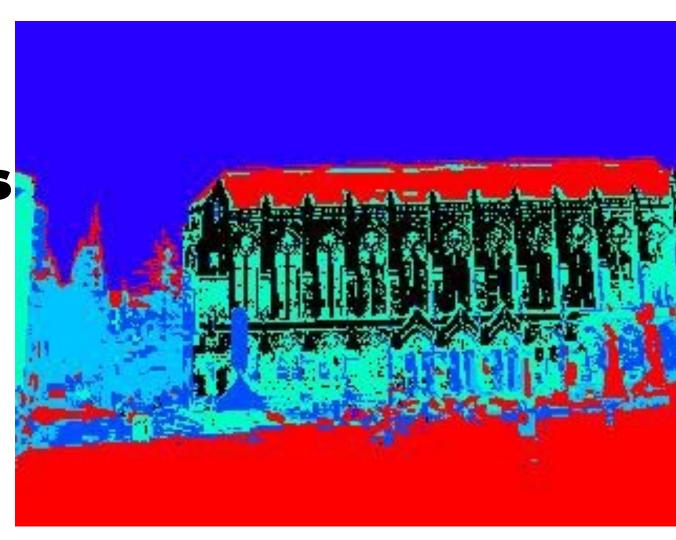


- Mid-Level: Image to feature (classical segmentation, grouping...)
 - What's the criterion? Gestalt psychologists suggest an intermediate vision stage whose underlying processes are grouping mechanisms, which are essential for separating objects from background. Certain "commonsense" principles, such as closure, symmetry, or similarity guide how to group pieces of image and locate boundary.



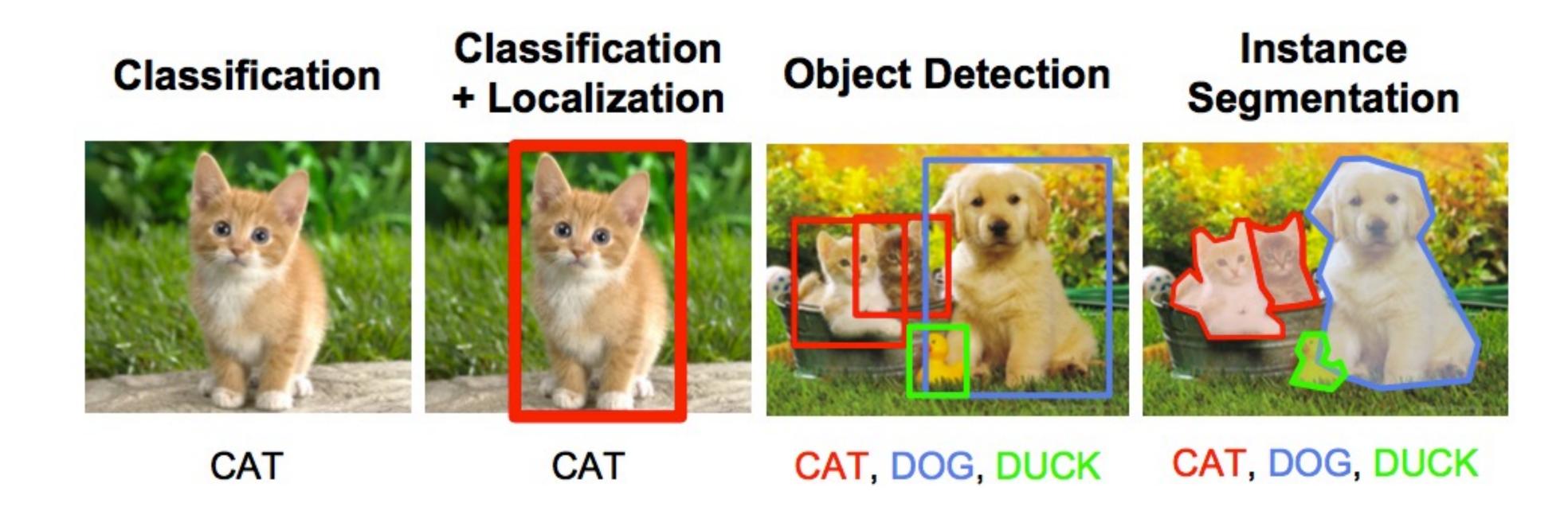
Clustering +
connected
component analysis

Object Structure



Three Stages in Computer Vision

- High-Level: Image to analysis (recognition, detection, semantic segmentation ...)
 - Facilitating semantic interpretation of visual data, and required for numerous applications like robotics, driver assistance, multi-media retrieval, biometrics and surveillance ...



Three Levels: An Example



"There's an edge!"

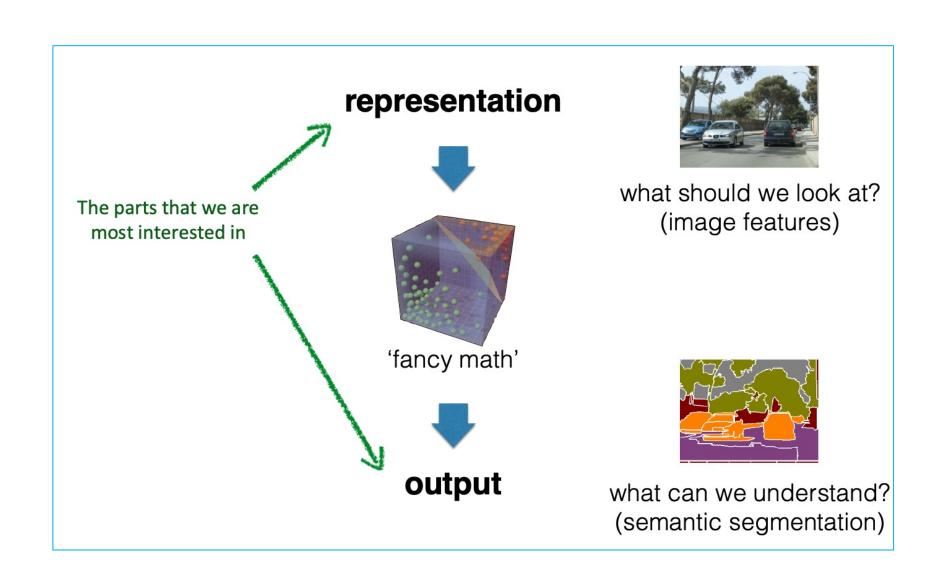


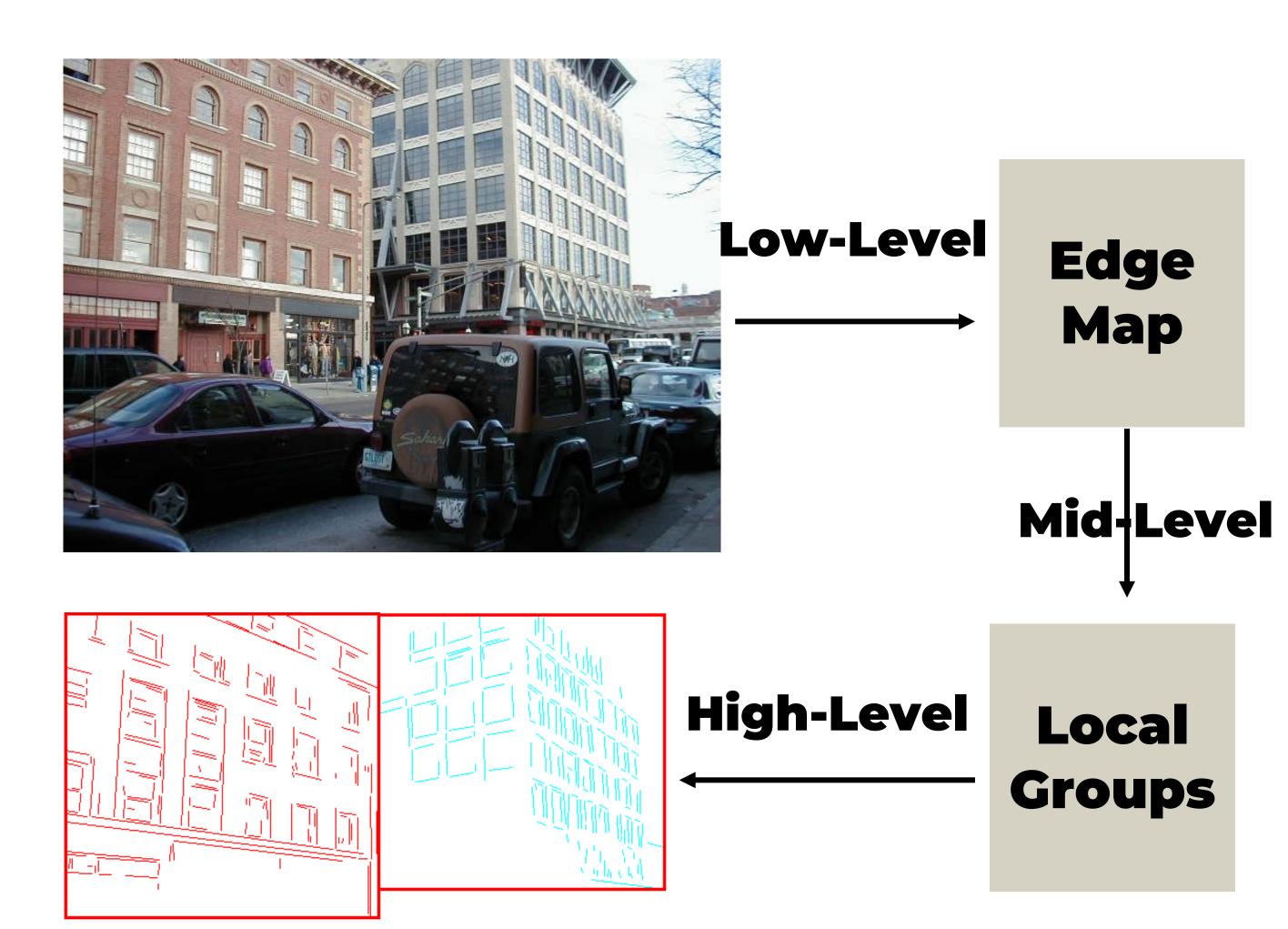
"There's an object and a background!"



"There's a chair!"

Example: A Simple Computer Vision Pipeline (1990s)





Building Recognition

Overview of Class Structure & Agenda

- Section 1 (8/23 9/06): Neuroscience, cognitive, and signal processing foundations of CV
- Section 2 (9/08 9/22): Extracting "good" features from 2D images (keyword: describe & match)
- Section 3 (9/27 10/25): From 2D to 3D vision (keyword: geometry & motion)
- Section 4 (10/27 11/15): Classical machine learning for CV tasks
- Section 5 (11/17 12/01): Modern deep learning for CV tasks

Lots of DSP and linear algebra await!

