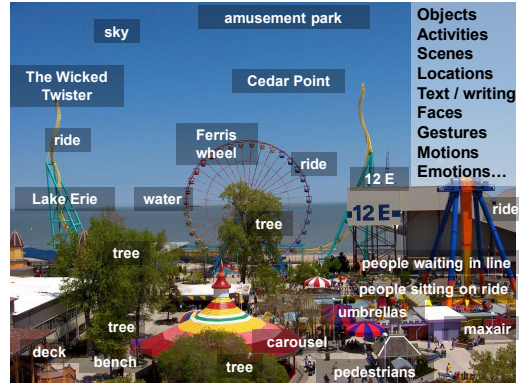


Computer Vision

- Automatic understanding of images and video
 1. Computing properties of the 3D world from visual data (*measurement*)
 2. Algorithms and representations to allow a machine to recognize objects, people, scenes, and activities. (*perception and interpretation*)

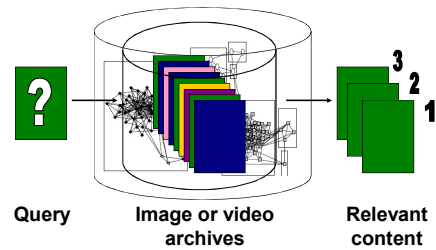
2. Vision for perception, interpretation



Computer Vision

- Automatic understanding of images and video
 1. Computing properties of the 3D world from visual data (*measurement*)
 2. Algorithms and representations to allow a machine to recognize objects, people, scenes, and activities. (*perception and interpretation*)
 3. Algorithms to mine, search, and interact with visual data (*search and organization*)

3. Visual search, organization

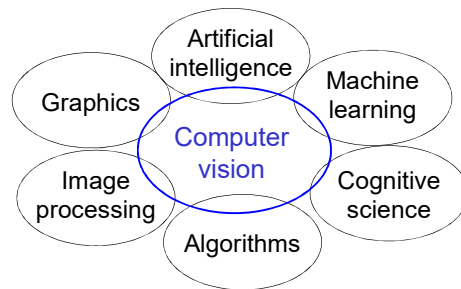


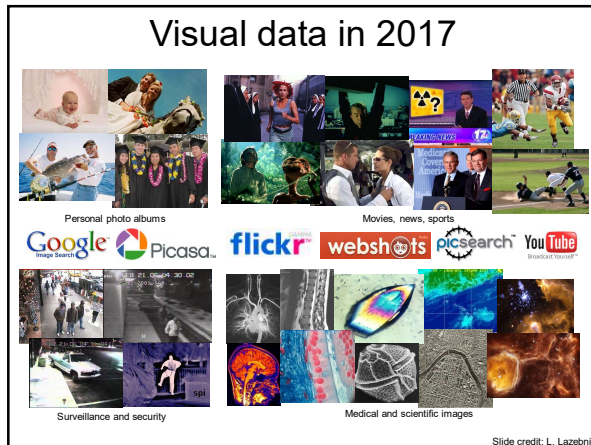
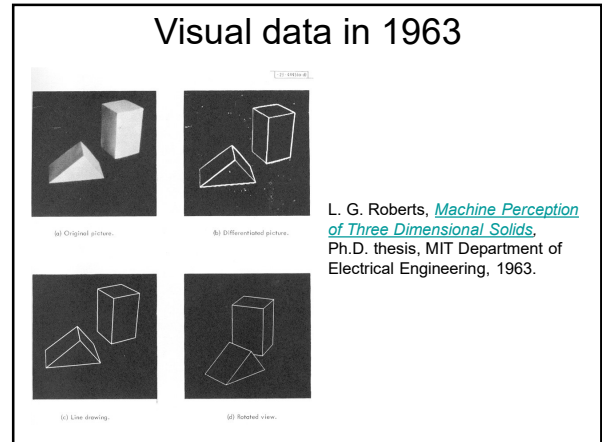
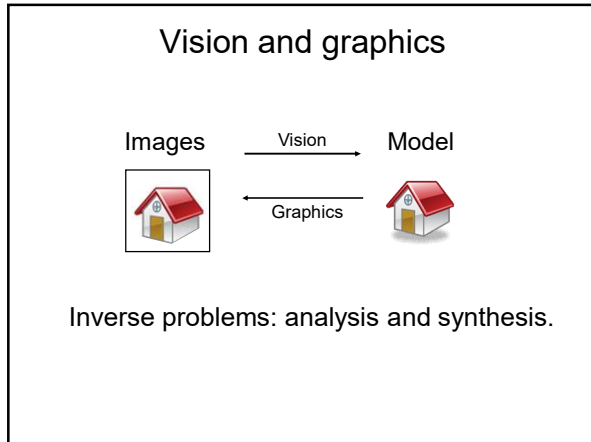
Computer Vision

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Course focus

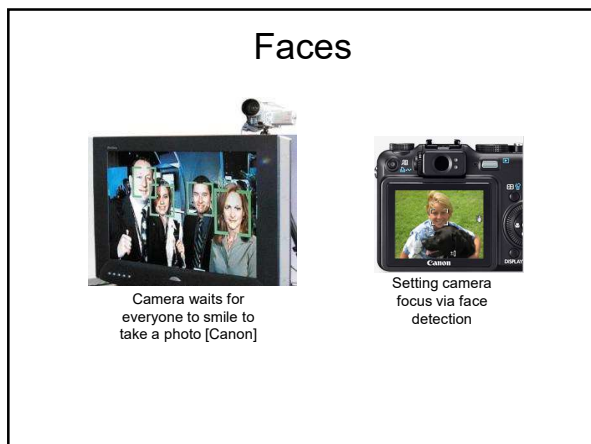
Related disciplines





Why recognition?

- Recognition a fundamental part of perception
 - e.g., robots, autonomous agents
- Organize and give access to visual content
 - Connect to information
 - Detect trends and themes
- Why now?



What else?

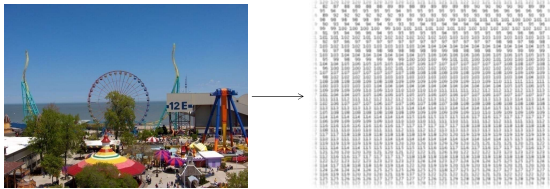
Obstacles?

MASSACHUSETTS INSTITUTE OF TECHNOLOGY
PROJECT MAC
Artificial Intelligence Group July 7, 1966
Vision Memo. No. 100.

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".

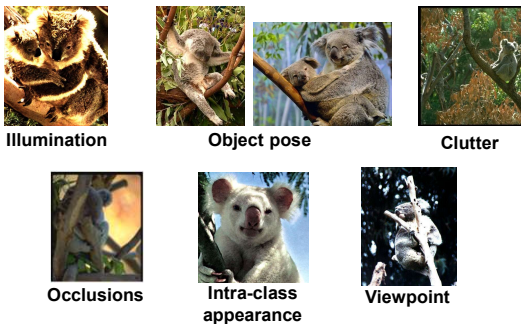
What the computer gets



Why is vision difficult?

- Ill-posed problem: real world much more complex than what we can measure in images
 - 3D → 2D
- Impossible to literally "invert" image formation process

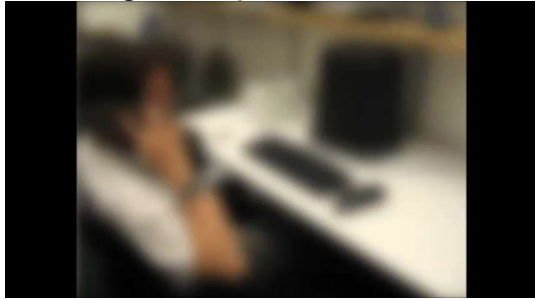
Challenges: many nuisance parameters



Challenges: intra-class variation



Challenges: importance of context



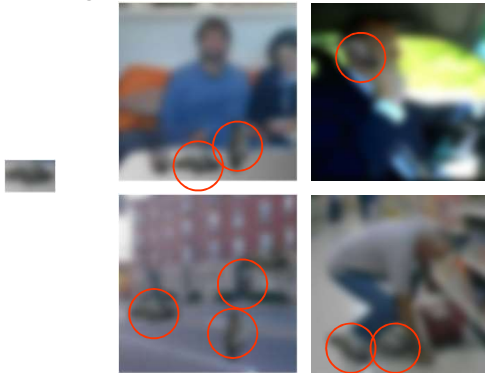
Video credit: Rob Fergus and Antonio Torralba

Challenges: importance of context



Video credit: Rob Fergus and Antonio Torralba

Challenges: importance of context

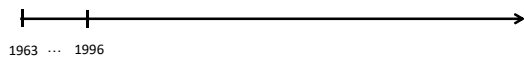
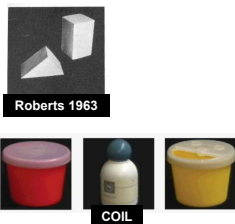


slide credit: Fei-Fei, Fergus & Torralba

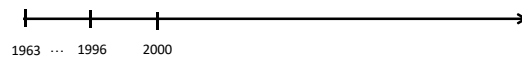
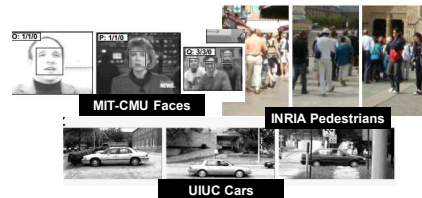
Challenges: complexity

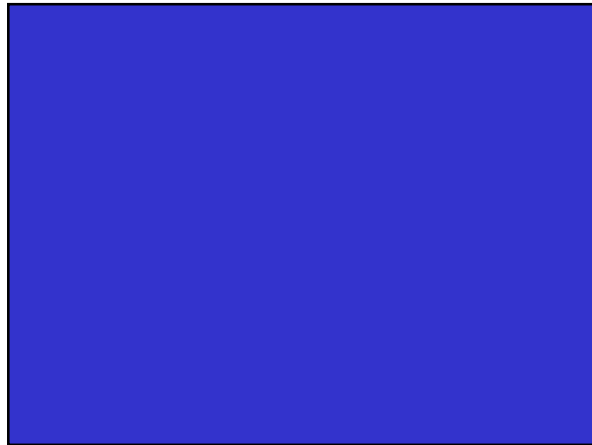
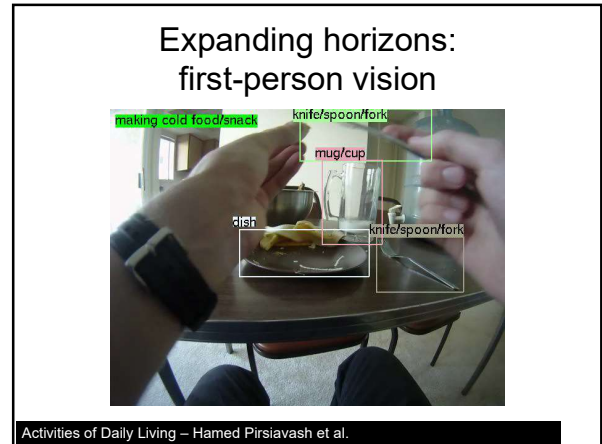
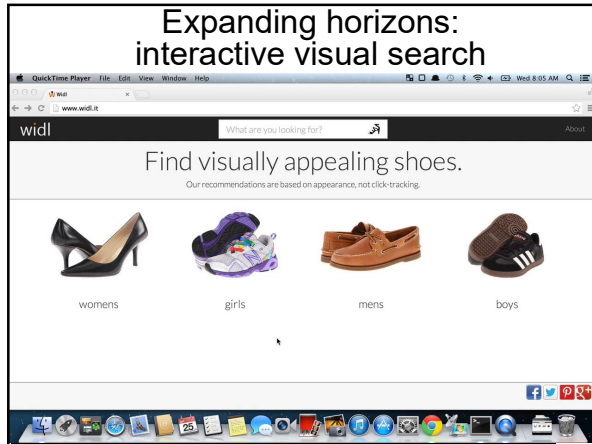
- Millions of pixels in an image
- 30,000 human recognizable object categories
- 30+ degrees of freedom in the pose of articulated objects (humans)
- 300+ hours of new video on YouTube per minute
- ...
- About half of the cerebral cortex in primates is devoted to processing visual information [Felleman and van Essen 1991]

Progress charted by datasets



Progress charted by datasets





This course

- Focus on current research in
 - Object recognition and categorization
 - Image/video retrieval, annotation
 - Some activity recognition
 - Related applications
- High-level vision and learning problems, innovative applications.

Goals

- Understand current approaches
- Analyze
- Identify interesting research questions
- Some hands-on experience

Prerequisites

- Courses in:
 - Computer vision
 - Machine learning
- Ability to analyze high-level conference papers

Basic format

- Early weeks (1-4):
 - Lectures by instructor
 - CNN tutorial
 - Paper reading
- Later weeks (5-11):
 - Paper discussion
 - Experiment
 - External paper presentation

Overview of requirements

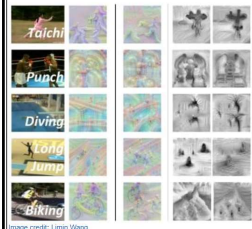
- **Discussions** will center on recent papers in the field
 - Write 2 paper reviews each week, due Mon
 - Serve as proponent/opponent ~twice
- **Student presentations**
 - Present an “external” from syllabus
 - Experiment on an assigned paper
- **2 implementation assignments**
- **Project with a partner**

Workload is fairly high

Assigned vs. external papers

Actions and objects in video

Detecting activities, actions, and events in images or video. Extracting foreground objects with video object segmentation. Recognition, relating actions to scenes, video descriptors, interactions with objects.



- Assigned**
- HICO: A Benchmark for Recognizing Human-Object Interactions in Images. Y-W. Chao, Z. Wang, Y. He, J. Wang, J. Deng. ICCV 2015. [pdf] [data]
 - Temporal Segment Networks: Towards Good Practices for Deep Action Recognition. L. Wang, Y. Xiong, Z. Wang, Y. Qiao, D. Lin, X. Tang, L. Van Gool. ECCV 2016. [pdf] [code]
 - Unsupervised Learning from Narrated Instruction Videos. J. Alayrac, P. Bojanowski, N. Agrawal, J. Sivic, I. Laptev, S. Lacoste-Julien. CVPR 2016. [pdf] [code] [video] [slides]
 - Learning to Segment Moving Objects in Videos. K. Fragkiadaki, P. Arbelaez, P. Felsen, J. Malik. CVPR 2015. [pdf] [code]
 - Action Recognition with Improved Trajectories. H. Wang and C. Schmid. ICCV 2013. [pdf] [web] [code] [IJCIV]
- External**
- Dynamic Image Networks for Action Recognition. Fernando, E. Gavves, A. Vedaldi, S. Gould. [pdf] [code]
 - Actions-Transformations. X. Wang, A. Farhadi, and A. Gupta. CVPR 2016. [pdf] [data]
 - Modeling actions through state changes. A Fathi and J Rehg. CVPR 2013. [pdf]

For inquiring minds

<http://vision.cs.utexas.edu/381V-fall2017>

Paper reviews

- Each week, review **two** of the assigned papers.
- Separately, summarize 2-3 “discussion points”
- Post each separately to Piazza following instructions on course “requirements” page.
- **Skip** reviews the week(s) you are presenting an external paper or experiment.

Paper review guidelines

- Brief (2-3 sentences) summary
- Main contribution
- Strengths? Weaknesses?
- How convincing are the experiments? Suggestions to improve them?
- Extensions? What’s inspiring?
- Additional comments, unclear points
- Relationships observed between the papers we are reading
- **due 8 pm Monday on Piazza**

Discussion point guidelines

- ~2-3 sentences/bullets per reviewed paper
- Recap of salient parts of your reviews
 - Key observations, lingering questions, interesting connections, etc.
- Will be shared to our class via Piazza
- Discussion points required for each class session (**due 8 pm Monday**)
- All encouraged to browse and post before and after class

External paper(s) presentation guidelines

- Well-organized talk that introduces it to the class
- About 15 minutes
- What to cover?
 - Problem overview, motivation
 - Algorithm explanation, technical details
 - Results summary
 - Relation to assigned reading where relevant
 - Demos, videos, other visuals etc. from authors
- See class webpage for more details.

Experiment guidelines

- Implement/download code for a main idea in the paper and show us toy examples:
 - Show (on a small scale) an example to analyze a strength/weakness of the approach
 - Experiment with different types of thoughtfully chosen data
 - Compare some aspect of assigned papers
- Key to a good experiment:
 - Don't duplicate what we saw in the paper!
 - **Not necessary to run whole thing end to end – focus, essentials**
- Present in class – about 20 minutes.
 - Don't recap the paper beyond 1-2 slides
- Include links to any tools or data in slides

Timetable and prep

- For external paper or experiment presentation, by the Wednesday **the week before** your presentation is scheduled:
 - Email draft slides to me
 - I'll provide feedback within the next few days
 - Hard deadline: 5 points per day late
- Please **coordinate with other** presenters in advance for your day to avoid duplication of papers
- Please **bring slides** on own laptop and check it prior to class
- Please **email me final slides** pdf after class session <lastname>_paper.pdf / <lastname>_expt.pdf

Projects

Possibilities:

- Extend a technique studied in class
- Analysis and empirical evaluation of an existing technique
- Comparison between two approaches
- Design and evaluate a novel approach
- Work in pairs
- Project proposal due mid-term

Important dates

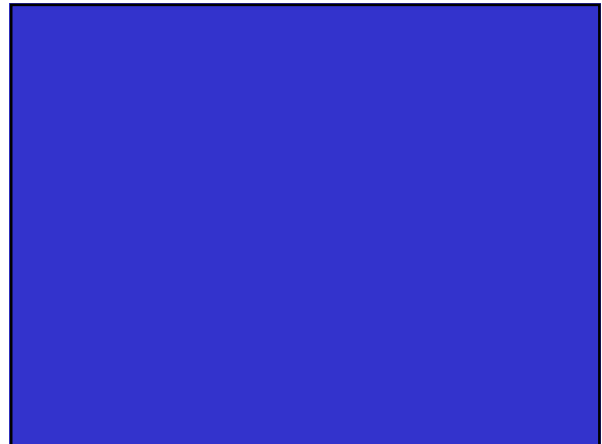
- Monday, Sept 4: paper topic preferences due to TA
- Monday, Sept 4: first set of 2 reviews due on Piazza
- Friday, Sept 22: first coding assignment due
- Wednesday, Oct 11: second coding assignment due
- Friday, Oct 13: second coding assignment follow-up run due
- Wednesday, Oct 25: project proposal due
- TBD in late Nov: poster printing deadline, 12 pm
- Wednesday, Dec 6: poster session in class, 1-4 pm
- Friday, Dec 8: final papers due

Grades

- Grades will be determined as follows:
 - 25% **participation** (includes attendance, in-class discussions, paper reviews)
 - 15% **coding** assignments
 - 35% **presentations** (includes drafts submitted one week prior, and in-class presentation)
 - 25% **final project** (includes proposal, poster, final paper)

Miscellaneous

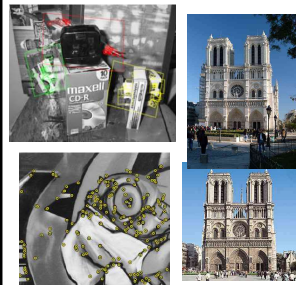
- Feedback welcome and useful!
- Slides on class website
- Discussion including assignment questions on Piazza
- No laptops, phones, etc. open in class please.
- Course is restricted to registered students



Syllabus Tour

- Learning objects and image representations
 - Instance recognition
 - Category recognition/detection
 - Self-supervised representation learning
 - ConvNet implementation tutorial
- Recognition on the move
 - Actions and objects in video
 - First-person vision
 - Embodied visual perception
- Potpourri
 - People
 - Visual data mining and discovery
 - Where to look
 - Language and vision

Instance recognition



Local invariant features, detection and description

Matching models to images

Indexing specific objects with bag-of-words descriptors

Category recognition/detection



Recognition as an image classification problem

Discriminative methods

Image descriptors

Convolutional neural networks

Benchmark datasets

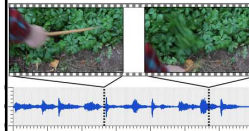
Object detection



Self-supervised representation learning



Unsupervised feature learning from "free" side information



(tracks in video, spatial layout in images, audio, colorization, ego-motion...)



CNN tutorial

Syllabus Tour

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Actions and objects in video

Detecting activities, actions, and events in images or video.

Video descriptors, interactions with objects and scenes.

Video object segmentation

First-person vision

Egocentric wearable cameras

Actions w/manipulated objects

Forecasting future activities

Developmental learning lessons

Embodied visual perception

Learning how to move for recognition, manipulation.

3D objects and next best view

Visual learning grounded in action and physical interaction

Visual recognition for robotics

Affordances

Syllabus Tour

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People



Human body pose
 Faces
 Fashion/clothing
 Attributes

Visual data mining and discovery

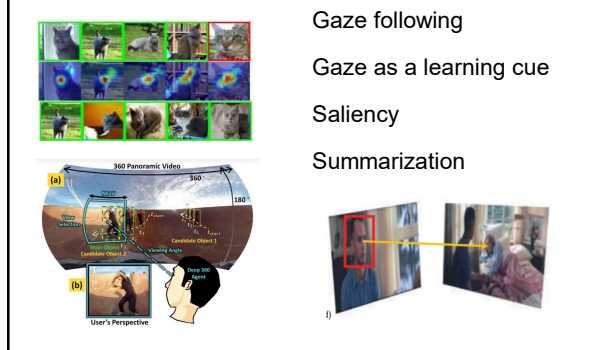


Discovering visual patterns in large-scale community photo collections

StreetView, Flickr data

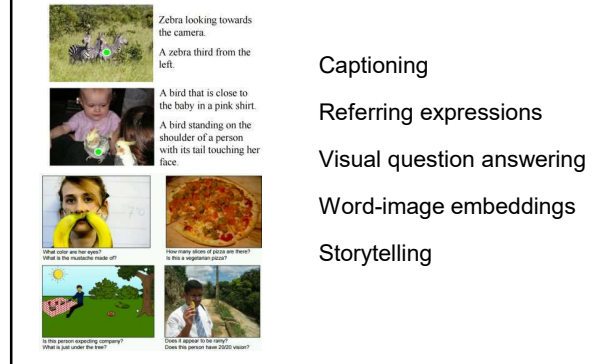
Demographics, geography, ecology, brands, fashion.

Where to look



Gaze following
 Gaze as a learning cue
 Saliency
 Summarization

Language and vision



Zebra looking towards the camera.
 A zebra third from the left.

A bird that is close to the baby in a pink shirt.
 A bird standing on the shoulder of a person with its tail touching her face.

What color are her eyes?
 What is the probability he'll get up?

How many slices of pizza are there?
 Is this a vegetable pizza?

Is this person operating a company?
 What is his name?
 Does it appear to be snow?
 Does this person have 2005 vision?

Captioning
 Referring expressions
 Visual question answering
 Word-image embeddings
 Storytelling

Not covered

- Low-level image processing
- Basic machine learning methods
- I will assume you already know these, or are willing to pick them up on your own.

Coming up

- Please read over course requirements online
- Due Monday 8 PM
 - Reading and paper reviews/discussion point posts for instance recognition
 - 6 top topic preferences to Wei-Lin via email