

Computer Vision

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

What does recognition involve?



Slide by Fei-Fei Li

Detection: are there people?



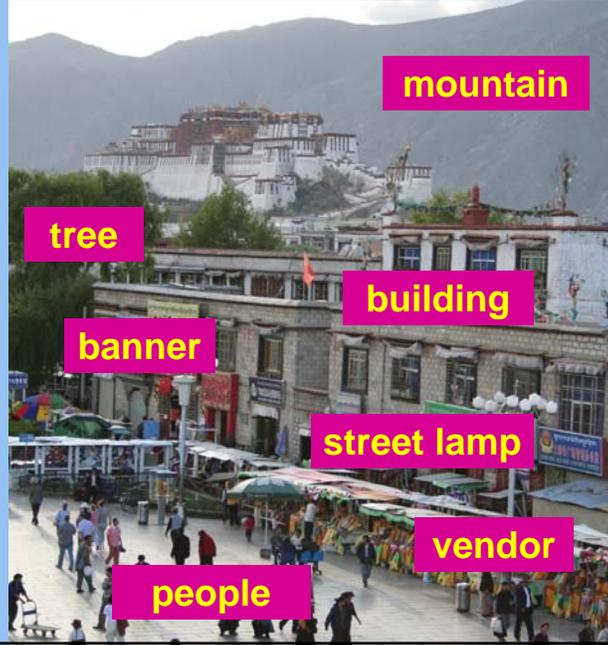
Slide by Fei-Fei Li

Activity: What are they doing?



Slide by Fei-Fei Li

Object categorization

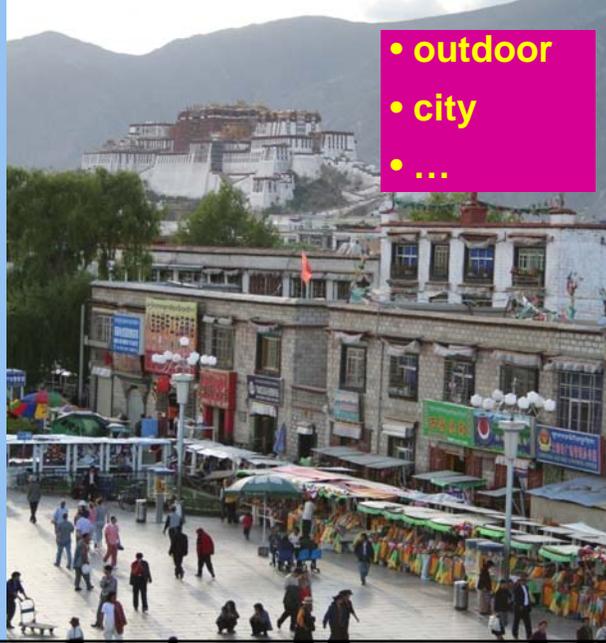


Slide by Fei-Fei Li

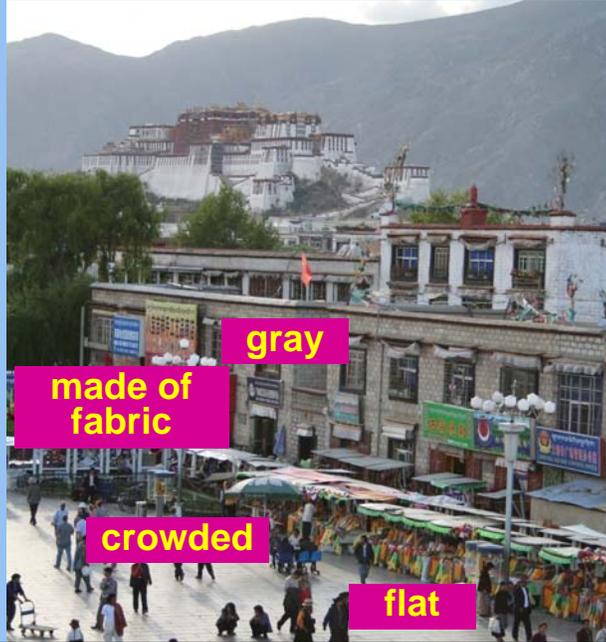
Instance recognition



Scene and context categorization



Attribute recognition



Object Categorization

- Task Description
 - “Given a small number of training images of a category, recognize a-priori unknown instances of that category and assign the correct category label.”
- Which categories are feasible visually?



“Fido” German shepherd dog animal living being

K. Grauman, B. Leibe

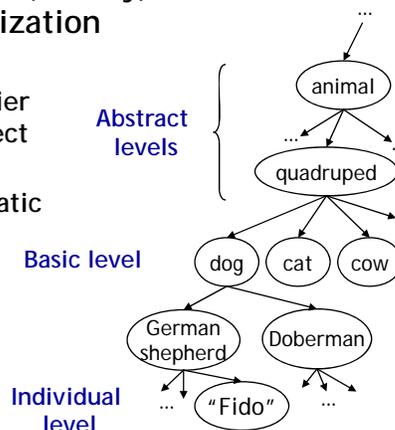
Visual Object Categories

- Basic Level Categories in human categorization [Rosch 76, Lakoff 87]
 - The highest level at which category members have similar perceived shape
 - The highest level at which a single mental image reflects the entire category
 - The level at which human subjects are usually fastest at identifying category members
 - The first level named and understood by children
 - The highest level at which a person uses similar motor actions for interaction with category members

K. Grauman, B. Leibe

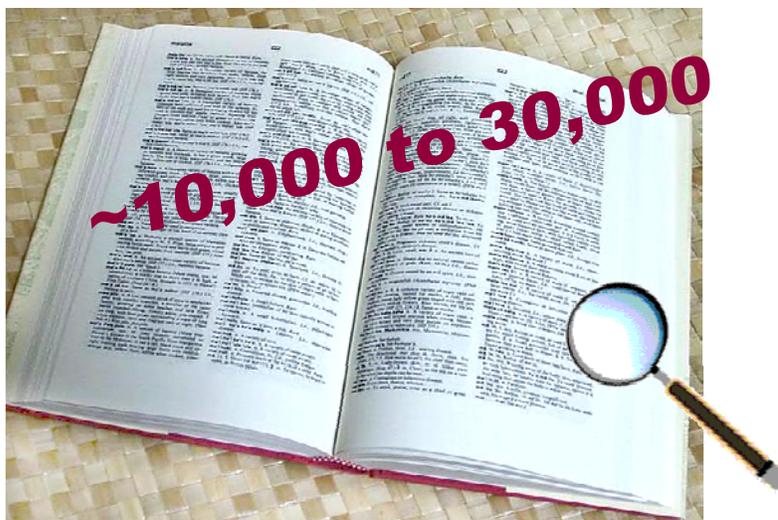
Visual Object Categories

- Basic-level categories in humans seem to be defined predominantly visually.
- There is evidence that humans (usually) start with basic-level categorization *before* doing identification.
 - ⇒ Basic-level categorization is easier and faster for humans than object identification!
 - ⇒ How does this transfer to automatic classification algorithms?



K. Grauman, B. Leibe

How many object categories are there?



Source: Fei-Fei Li, Rob Fergus, Antonio Torralba.

Biederman 1987



Other Types of Categories

- Functional Categories
 - e.g. chairs = "something you can sit on"



K. Grauman, B. Leibe

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?

Autonomous agents able to detect objects



<http://www.darpa.mil/grandchallenge/gallery.asp>

Posing visual queries



Yeh et al., MIT



Digital Field Guides Eliminate the Guesswork

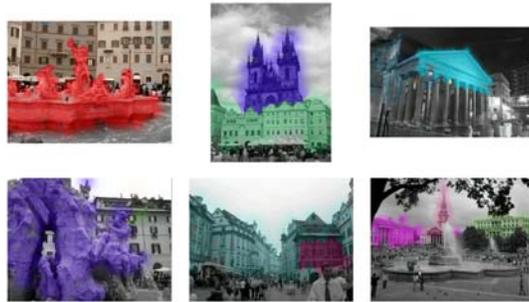
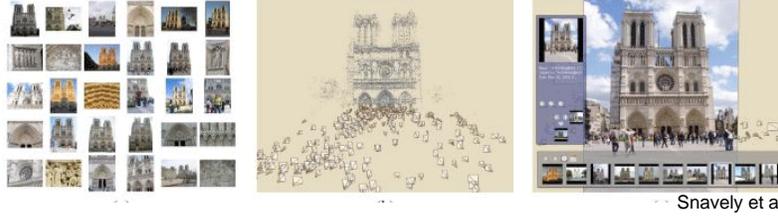


Belhumeur et al.

Kooaba, Bay & Quack et al.

Finding visually similar objects

Exploring community photo collections



Simon & Seitz

Discovering visual patterns



Objects Sivic & Zisserman



Categories Lee & Grauman



Actions Wang et al.

Auto-annotation



Figure 9. Results of automatic object-level annotation with bounding boxes. Groundtruth annotation is shown with dashed lines, correct detection with solid green lines, false detections with solid red lines. Auto-annotation with related Wikipedia articles is also shown. All results are also labeled with their GPS position and estimated tags (not shown here).

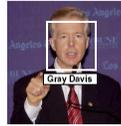
Gammeter et al.



President George W. Bush makes a statement in the Rose Garden while Secretary of Defense Donald Rumsfeld looks on, July 23, 2003. Rumsfeld said the United States would release graphic photographs of the dead sons of Saddam Hussein to prove they were killed by American troops. Photo by Larry Downing/Routers



British director Sam Mendes and his partner actress Kate Winslet arrive at the London premiere of 'The Road to Iliodon', September 18, 2002. The film stars Tom Hanks as a Chicago hit man who has a separate family life and co-stars Paul Newman and Jude Law. REUTERS/Dan Chung



Incumbent California Gov. Gray Davis (news - web sites) leads Republican challenger Bill Simon by 10 percentage points - although 17 percent of voters are still undecided, according to a poll released October 22, 2002 by the Public Policy Institute of California. Davis is shown speaking to reporters after his debate with Simon in Los Angeles, on Oct. 7. (Jim Raymon/Routers)

T. Berg et al.

Challenges

Challenges: robustness



Illumination



Object pose



Clutter



Occlusions



Intra-class appearance



Viewpoint

Challenges: context and human experience



Context cues

Challenges: context and human experience



Context cues



Function

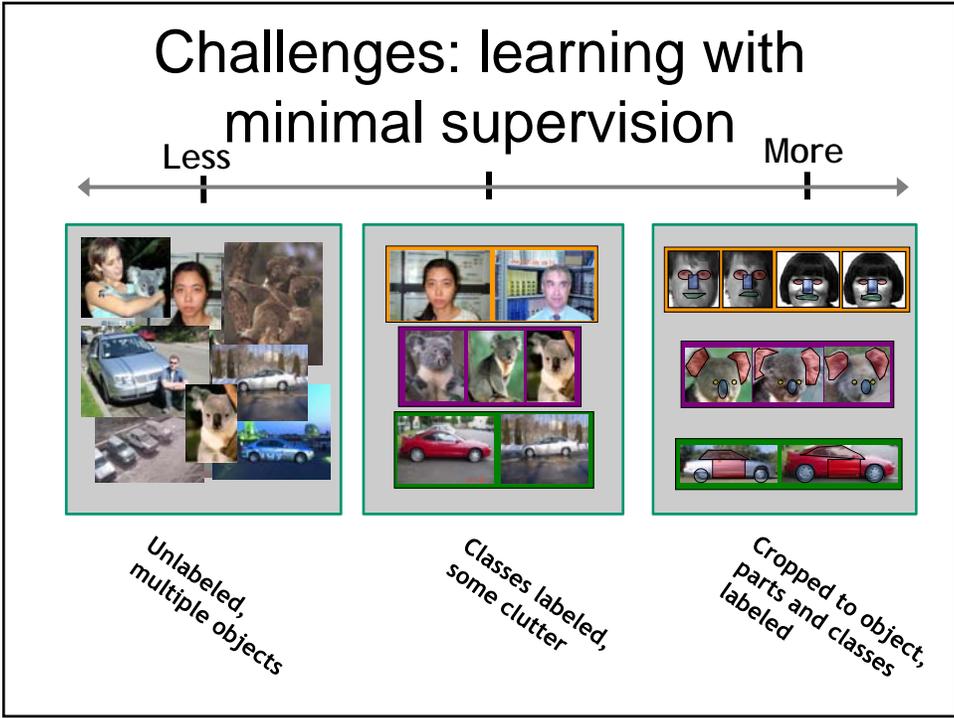


Dynamics

Video credit: J. Davis

Challenges: scale, efficiency

- Half of the cerebral cortex in primates is devoted to processing visual information
- ~20 hours of video added to YouTube per minute
- ~5,000 new tagged photos added to Flickr per minute
- Thousands to millions of pixels in an image
- 30+ degrees of freedom in the pose of articulated objects (humans)
- 3,000-30,000 human recognizable object categories



What kinds of things work best today?

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

Reading license plates,
zip codes, checks



Recognizing flat, textured
objects (like books, CD
covers, posters)

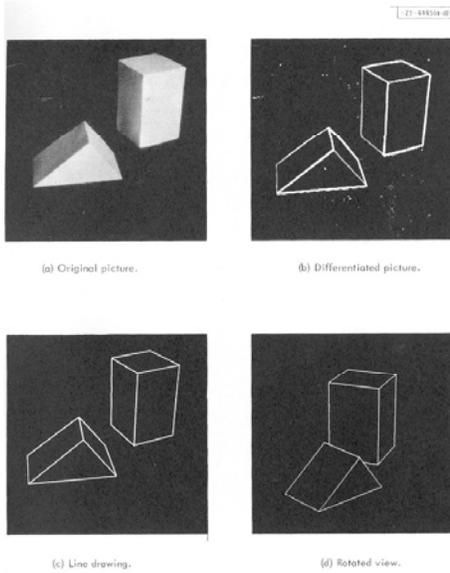


Frontal face detection



Fingerprint recognition

Inputs in 1963...



L. G. Roberts, *Machine Perception of Three Dimensional Solids*, Ph.D. thesis, MIT Department of Electrical Engineering, 1963.

... and inputs today



Slide credit: L. Lazebnik

... and inputs today



introductions



This course

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

Goals

- Understand current approaches
- Analyze
- Identify interesting research questions

Expectations

- **Discussions** will center on recent papers in the field
 - Paper reviews each week
- **Student presentations**
 - Papers and background reading
 - Experiment presentation
- **2 implementation assignments**
- **Project**

Workload is fairly high

Prerequisites

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

Paper reviews

- Each week, review two of the assigned papers.
- Email me and TA by Thurs 9 PM
- Skip reviews the week(s) you are presenting.

Paper review guidelines

- Brief (2-3 sentences) summary
- Main contribution
- Strengths? Weaknesses?
- How convincing are the experiments?
Suggestions to improve them?
- Extensions?
- Additional comments, unclear points
- Relationships observed between the papers we are reading

Paper presentation guidelines

- Read 3 selected papers in topic area
- Well-organized talk, about 30-45 minutes
- What to cover?
 - Problem overview, motivation
 - Algorithm explanation, technical details
 - Any commonalities, important differences between techniques covered in the papers.
- See handout and class webpage for more details.

Experiment guidelines

- Implement/download code for a main idea in the paper and show us toy examples:
 - Experiment with different types of (mini) training/testing data sets
 - Evaluate sensitivity to important parameter settings
 - Show (on a small scale) an example to analyze a strength/weakness of the approach
- Present in class – about 30 minutes.
- Share links to any tools or data.

Timetable for presenters

- For papers or experiments, by the Friday **the week before** your presentation is scheduled:
 - Email draft slides to me, and schedule a time to meet, do dry run, discuss.
 - This is a hard deadline: 5 points off automatically per day late
- See course webpage for examples of good reviews, presentations.

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
 - Thorough survey / review paper
-
- Work in pairs, except for survey.

Miscellaneous

- Feedback welcome and useful
- No laptops, phones, etc. in class please
- Check class website
- I'll use Blackboard to email class



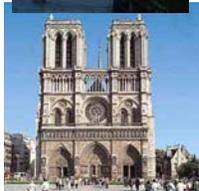
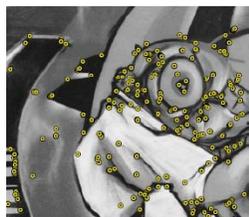
Syllabus tour

- I. Object recognition fundamentals
- II. Beyond modeling individual objects
- III. Human-centered recognition

Syllabus tour

- I. Object recognition fundamentals
 - A. Local features and matching object instances
 - B. Large-scale search and mining
 - C. Classification and detection of categories
 - D. Mid-level representations

Local features and matching object instances



Local invariant features,
detection and description

Matching models to
images

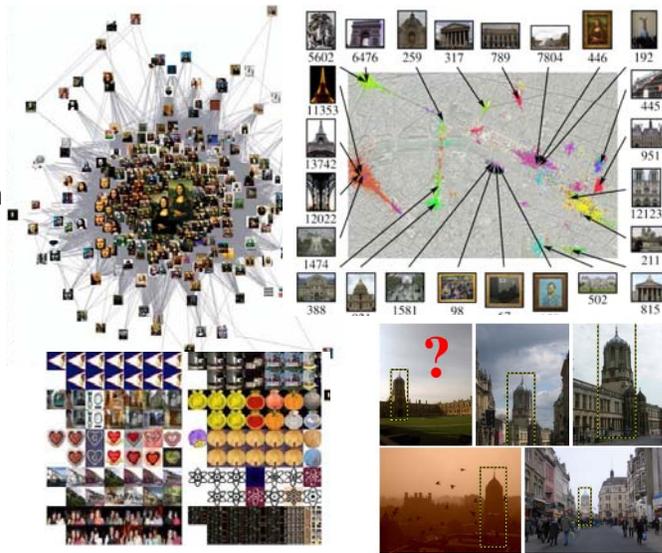
Indexing specific objects
with bag-of-words
descriptors

Large-scale image/object search and mining

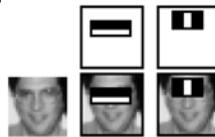
Using instance recognition for large-scale search

Scalable hashing algorithms

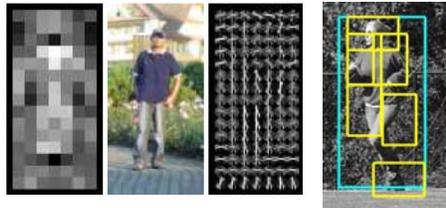
Adopting text retrieval insights



Classification and detection for object categories

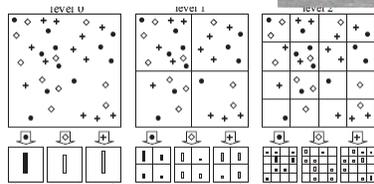


Detection as classification problem



Discriminative methods

Global representations with rigid spatial



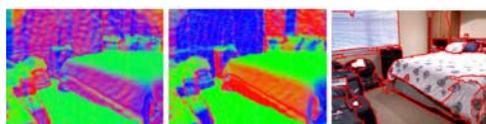
Faces and pedestrians as case studies

Mid-level representations



Segmentation

Category-independent
region ranking



Surface Normals

Aligned Normals

Segmentation

Surface estimation



Hypotheses

Syllabus tour

II. Beyond modeling individual objects

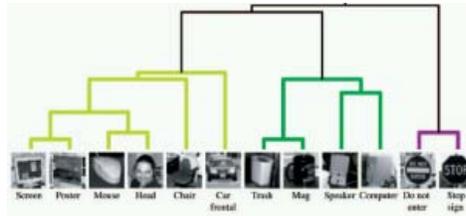
- A. Context and scenes
- B. Dealing with many categories
- C. Describing objects with attributes
- D. Importance and saliency

Context and scenes

The scene, the other objects, the spatial layout, geometry of surfaces --- all tell us more about what is reasonable to detect.



Dealing with many categories



Sharing features between classes

Transfer learning

Learning from few examples

Category hierarchies

Describing objects with attributes

Beyond naming object by category, we should be able to describe their properties, or use descriptions to understand novel objects.

Found 1344 results for **smiling asian men with glasses** in 0.220 s

Found | Faces | Images

(a) Smiling (b) ? (c) Not smiling

(d) Natural (e) ? (f) Manmade

Naming → Aeroplane

Description → Unknown
Has Wheel
Has Wood

Unusual attributes → Bird
No Head
No Beak

Unexpected attributes → Motorbike
Has Cloth

Textual description → Has Horn
Has leg
Has Head
Has Wool

Detailed description: This slide illustrates a search interface for 'smiling asian men with glasses' and its results. It shows a grid of images categorized by 'Faces' and 'Images'. The 'Faces' section includes three images labeled (a) Smiling, (b) ?, and (c) Not smiling. The 'Images' section includes three images labeled (d) Natural, (e) ?, and (f) Manmade. To the right, a vertical list of images is annotated with arrows pointing to descriptive labels: 'Naming' points to an aeroplane; 'Description' points to a rickshaw (Unknown, Has Wheel, Has Wood); 'Unusual attributes' points to a bird (Bird, No Head, No Beak); 'Unexpected attributes' points to a motorbike (Motorbike, Has Cloth); and 'Textual description' points to a sheep (Has Horn, Has leg, Has Head, Has Wool).

Saliency and importance

Among all items in the scene, which deserve attention (first)?
What makes images interesting or memorable?

a) Most memorable images (86%)

c) Least memorable images (34%)

Detailed description: This slide shows a grid of images used to study saliency and importance. The top row, labeled 'a) Most memorable images (86%)', contains 12 images: a red maple leaf, a red sports car, a woman's face, a blue cartoon character, a toilet, a person in a red shirt, and a group of colorful balloons. The bottom row, labeled 'c) Least memorable images (34%)', contains 12 images: a white car, a white car, a white car, a landscape, and a building. The images in row 'a' are more visually distinct and varied, while the images in row 'c' are more similar and less visually striking.

Syllabus tour

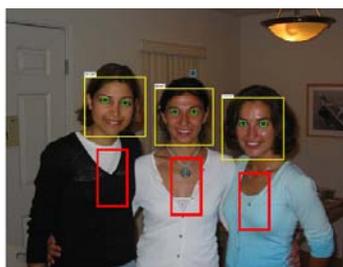
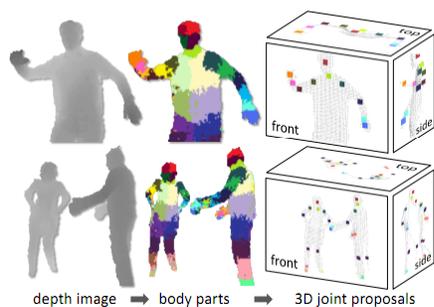
III. Human-centered recognition

- A. Pictures of people
- B. Activity recognition
- C. Egocentric cameras
- D. Human-in-the-loop interactive systems

Pictures of people

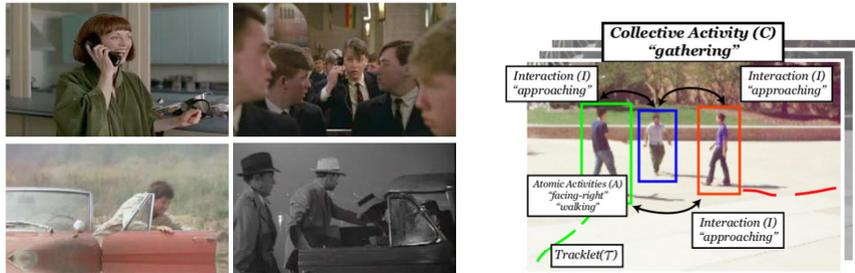
Finding people and their poses

Automatic face tagging



Activity recognition

Recognizing human actions in images and video



Egocentric cameras

Recognizing objects and actions from a first person point of view



Input: Egocentric video of the camera wearer's day



1:00 pm 2:00 pm 3:00 pm 4:00 pm 5:00 pm 6:00 pm

Output: Storyboard summary of important people and objects

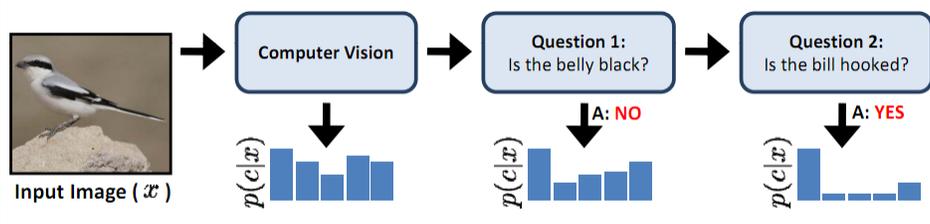
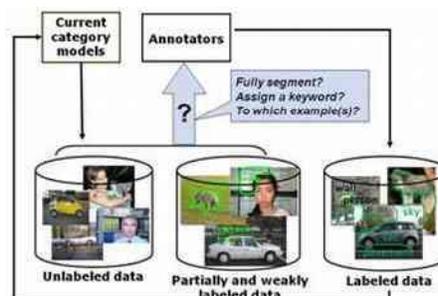


Human-in-the-loop interactive systems

Human-in-the-loop learning

Active annotation collection

Crowdsourcing



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

- Talk next Friday at 11:30 am in ACES 2.402:
Silvio Savarese, Univ. of Michigan
“Understanding the 3d world from images”
- Review syllabus, select 4 topic preferences
 - Email to Austin (TA) by Wed Sept 5 at 5 pm
- Read assigned papers for “local features and matching for object instances”, and review the Sivic and Lowe papers.