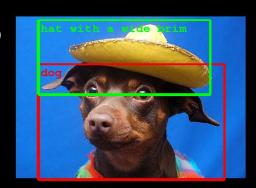
CS4495/6495 Introduction to Computer Vision

8A-L1 Introduction to recognition



What does recognition involve?



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

Verification: is that a lamp?



Detection: are there people?



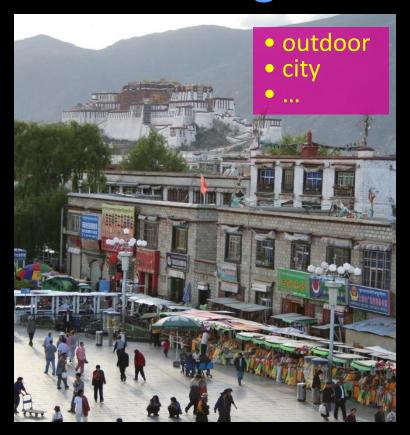
Identification: is that Potala Palace?



Object categorization



Scene and context categorization



Instance-level recognition problem



John's car

Generic categorization problem







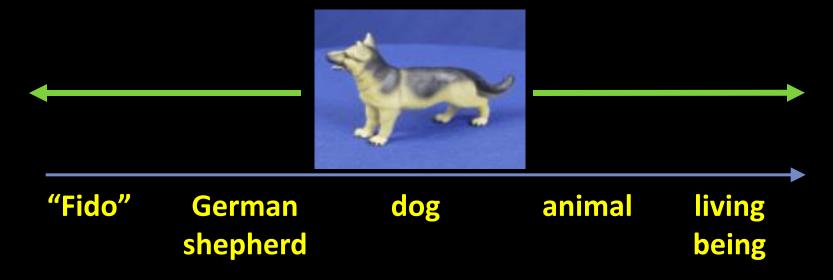




Object Categorization

Task: Given a (small) number of training images of a category, recognize a-priori unknown instances of that category and assign the correct category label.

Object Categorization



Which categories are the best for visual identification?

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

Visual Object Categories

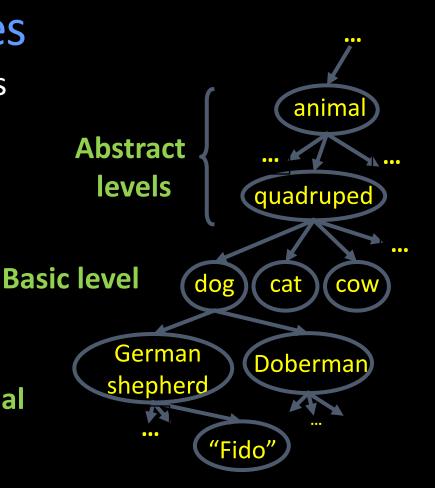
Basic Level Categories in human categorization [Rosch 76, Lakoff 87]

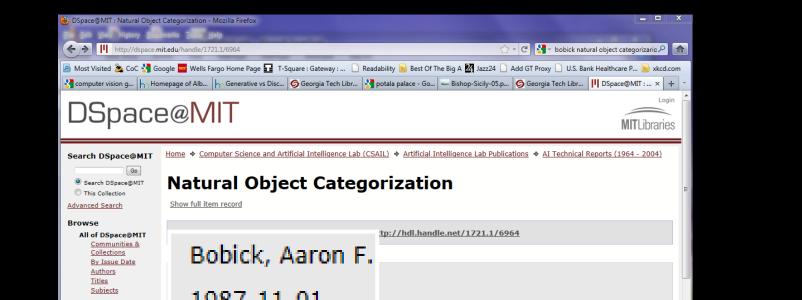
- The level at which human subjects are usually fastest at identifying category members
- The first level named and understood by children
- The highest level at which a person uses similar motor actions for interaction with category members

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.

Individual level





gorizing natural objects. To provide a criteria for categorization we propose that the

he inference of unobserved properties of objects from the observed properties.

claim about the structure of the world. We first define an evaluation function that measures how well a set of categories

supports the inference goals of the observer. Entropy measures for property uncertainty and category uncertainty are combined through a free parameter that reflects the goals of the observer. Natural categorizations are shown to be those that are stable with respect to this free parameter. The evaluation function is tested in the domain of leaves and is found to be sensitive to the structure of the natural categories corresponding to the different species. We next develop a categorization

paradigm that utilizes the categorization evaluation function in recovering natural categories. A statistical hypothesis

generation algorithm is presented that is shown to be an effective categorization procedure. Examples drawn from several

natural domains are presented, including data known to be a difficult test case for numerical categorization techniques. We next extend the categorization paradigm such that multiple levels of natural categories are recovered; by means of recursively invoking the categorization procedure both the genera and species are recovered in a population of anaerobic bacteria. Finally,

constructed in an arbitrary world, we present the Principle of Natural Modes as a

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Abstract:

a method is presented for evaluating the utility of features in recovering natural categories. This method also provides a

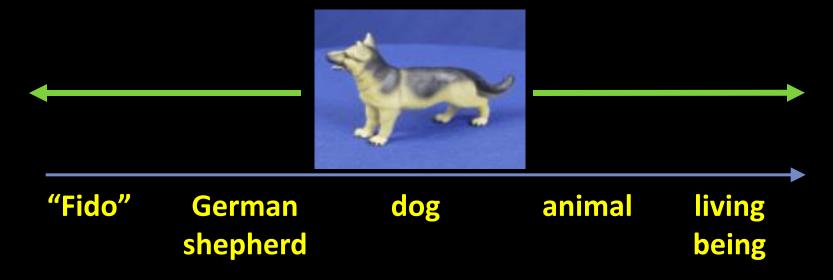
mechanism for determining which features are constrained by the different processes present in a multiple modal world. Next ↑ Previous Previous Highlight all Match case

Natural
Object
Categorization

Aaron F. Bobick

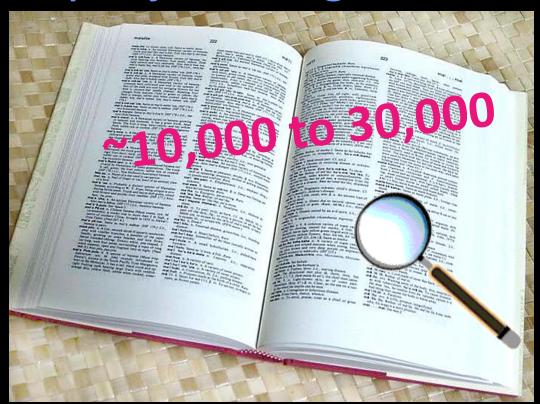
MIT Artificial Intelligence Laboratory

Object Categorization



Which categories are the best for visual identification?

How many object categories are there?



Other Types of Categories

Functional Categories

e.g. chairs = "something you can sit on"











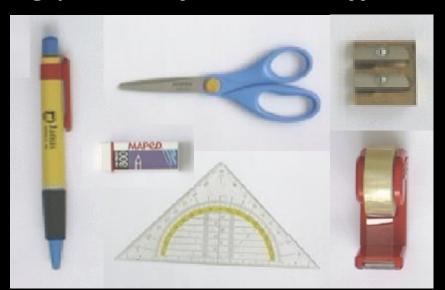






Other Types of Categories

Ad-hoc categories e.g. "something you can find in an office environment"



K. Grauman, B. Leibe

Autonomous agents able to detect objects





Labeling people

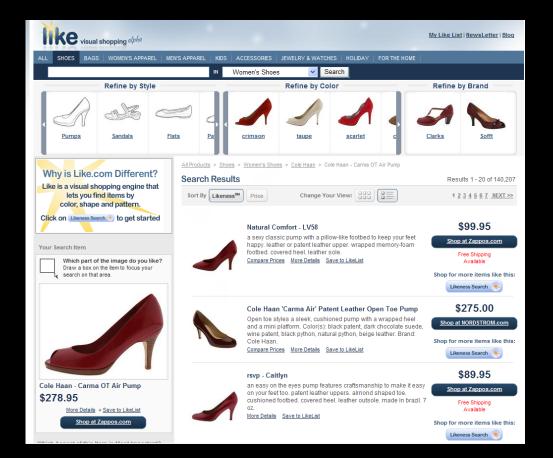


Posing visual queries

Digital Field Guides Eliminate the Guesswork



Finding visually similar objects



So why is this hard?

Challenges: Robustness



Illumination



eiect nose





Clutter

Challenges: Robustness



Occlusions



Intra-class appearance

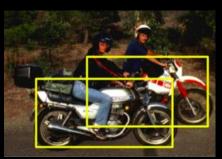


Viewpoint

Challenges: Robustness

Realistic scenes are crowded, cluttered, have overlapping objects.







Challenges: Importance of context



Fei-Fei, Fergus & Torralba

Challenges: Importance of context



Fei-Fei, Fergus & Torralba

Challenges: complexity

- Thousands to millions of pixels in an image
- 3,000-30,000 human recognizable object categories
- 30+ degrees of freedom in the pose of articulated objects (humans)

Challenges: complexity

- Billions of images indexed by Google Image Search
- In 2011, 6 billion photos uploaded per month
- Approx one billion million camera phones sold in 2013
- About half of the cerebral cortex in primates is devoted to processing visual information [Felleman and van Essen 1991]

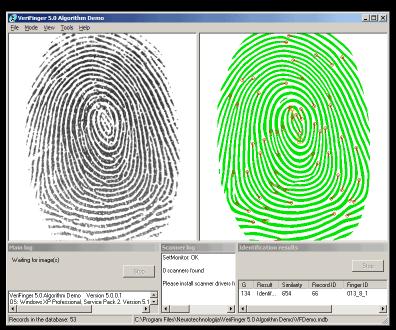
So what works?

 Reading license plates (real easy), zip codes, checks



Lana Lazebnik

- Reading license plates, zip codes, checks
- Fingerprint recognition



Lana Lazebnik

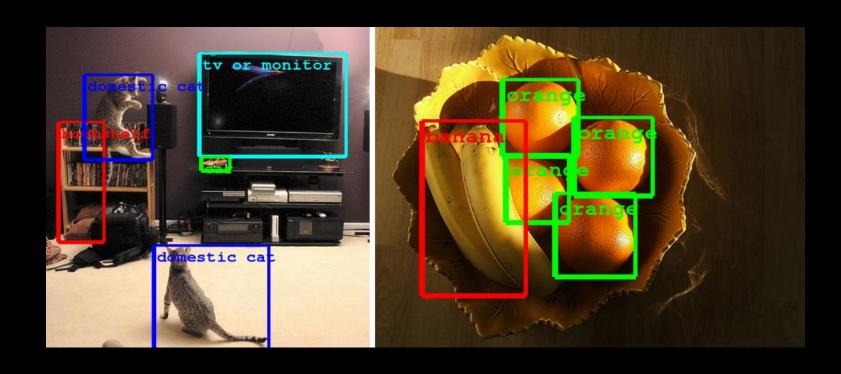
- Reading license plates, zip codes, checks
- Fingerprint recognition
- Face detection (Today recognition)



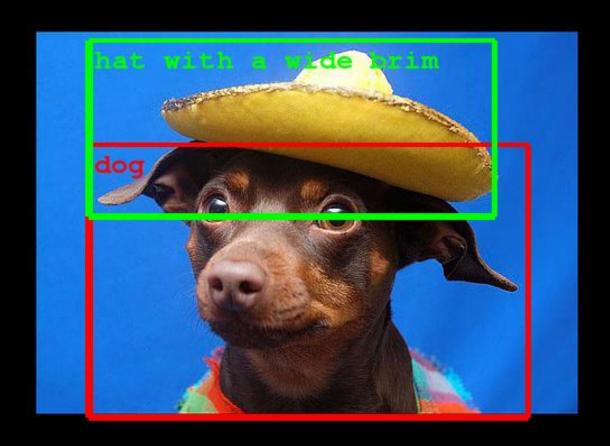
- Reading license plates, zip codes, checks
- Fingerprint recognition
- Face detection (Today recognition)
- Recognition of flat textured objects (CD covers, book covers, etc.)



Just in: GoogleNet 2014



Just in: GoogleNet – no context needed?



Going forward

- Much of strong label "recognition" is really machine learning applied to patterns of pixel intensities.
- We'll focus on some principles of generative vs discriminative methods and image representations that they use.
- Then we'll spend some time on activity recognition from video.