

CMPSCI 370: Intro. to Computer Vision

Introduction to recognition

University of Massachusetts, Amherst
March 29, 2014

Instructor: Subhransu Maji

Announcements

- Midterm has been graded
 - Average score: **54.2** (out of 80)
 - Come by my office hours
 - if you have any questions
 - or did not collect the midterm in class
 - or to chat about the latest AI technology (AlphaGo, Holoportation,)
- Homework 3 grades will be available shortly
- **No class** this Thursday (3/31) due to instructor's travel
- **No honors section** today

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Common mistakes ...

- What is a Bayer filter for?
 - “for image smoothing”
 - “for color sensing in digital cameras”
- A technique to enhance the contrast of an image:
 - “sharpen the image” — sharpening is not the same as contrast enhancement
 - “gamma/log-normalization”, “brightness stretching”, “histogram equalization”
- Factor that lead to edges
 - “gx, gy is high”
 - “occlusion, shadows, change in surface orientation, texture,...”

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Object Recognition: Overview and History



Slides adapted from Svetlana Lazebnik, Alex Berg, Fei-Fei Li, Rob Fergus, Antonio Torralba, and Jean Ponce

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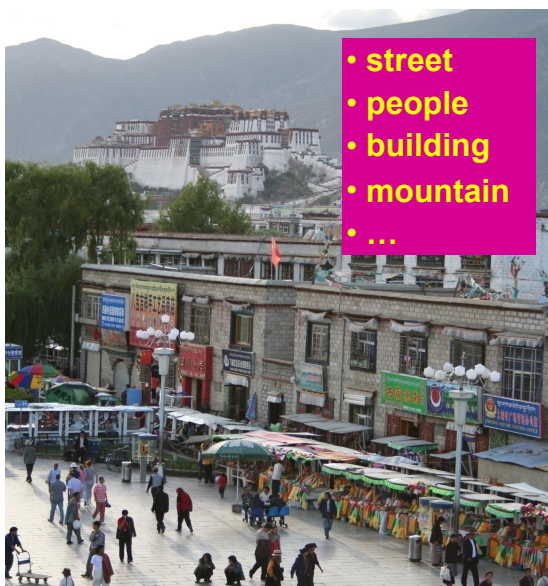
Scene categorization



- outdoor/indoor
- city/forest/factory/etc.

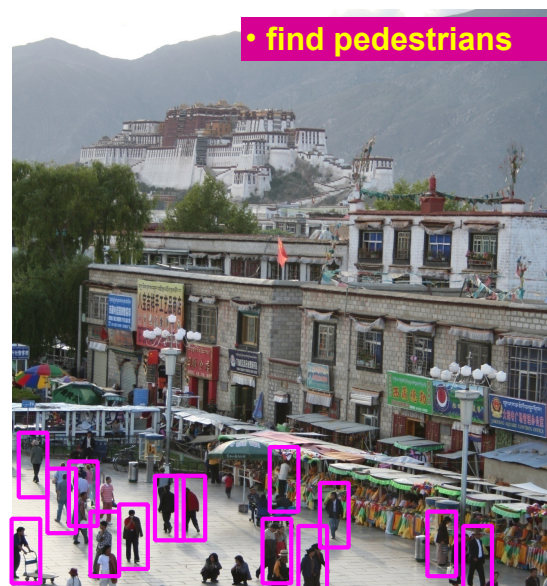
6

Image annotation/tagging



- street
- people
- building
- mountain
- ...

Object detection



- find pedestrians

Activity recognition

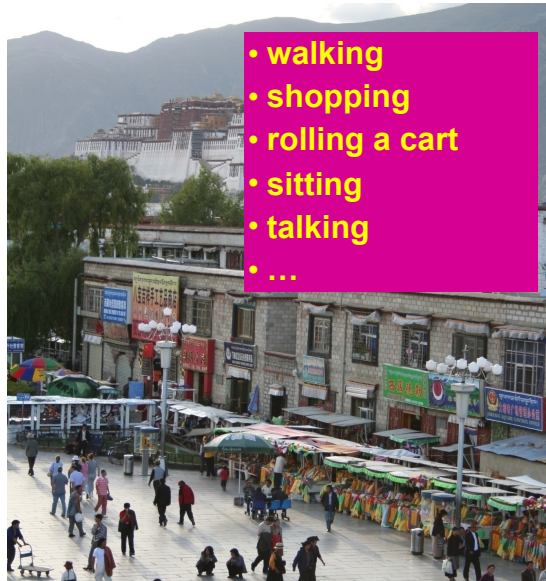


Image parsing

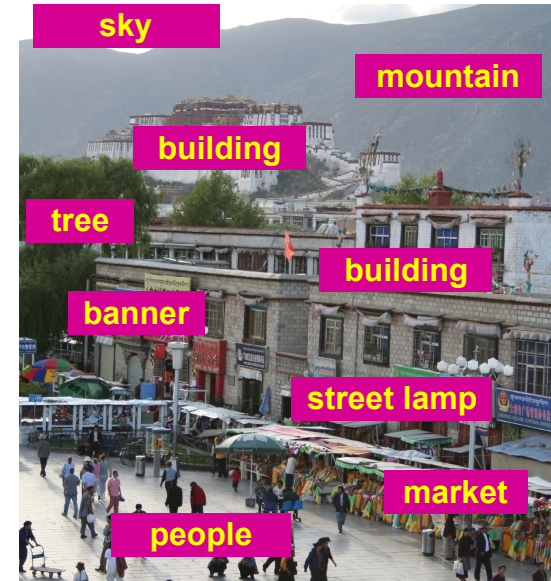
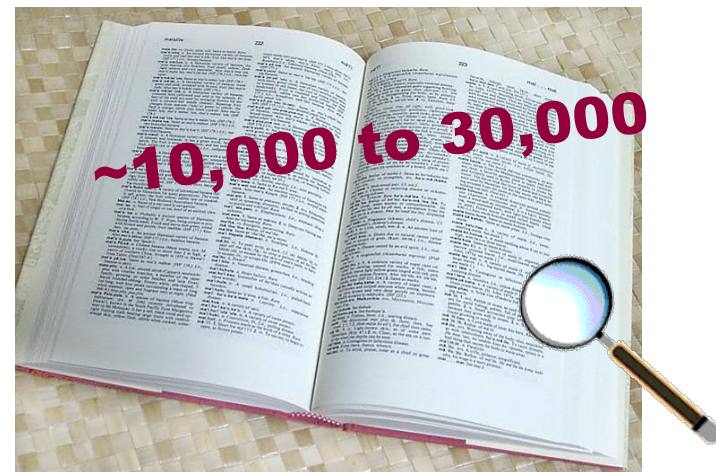


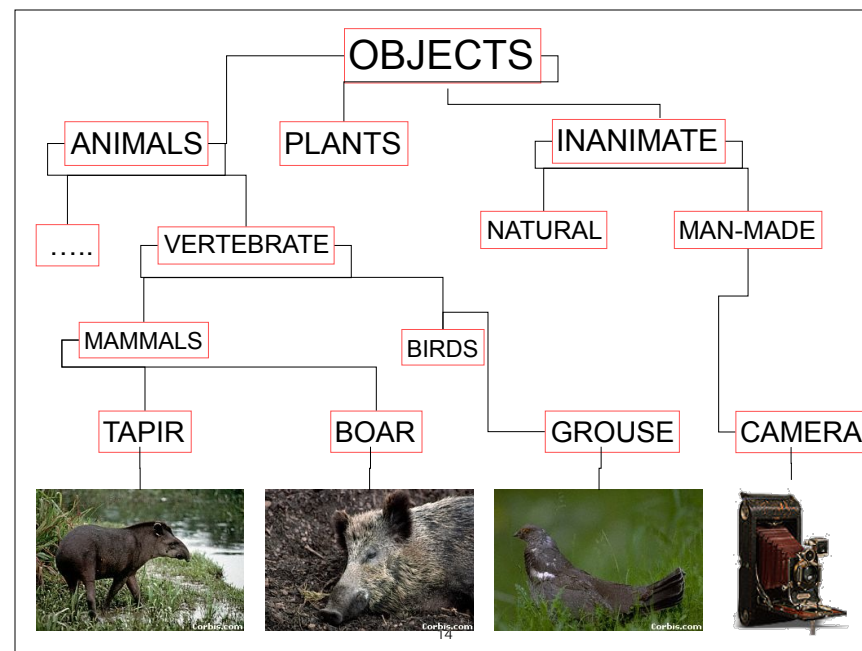
Image understanding?



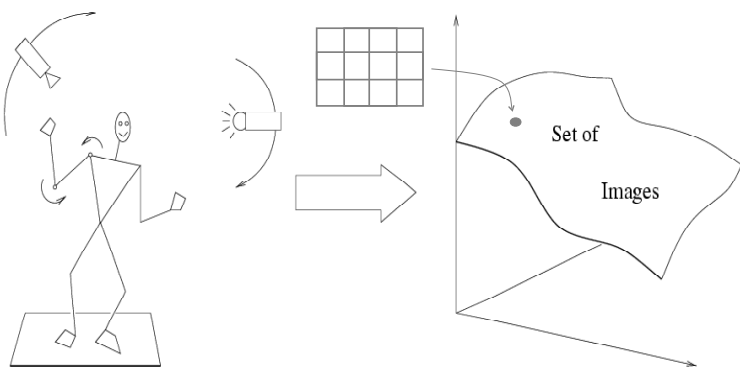
How many visual object categories are there?



[http://wexler.free.fr/library/files/biederman%20\(1987\)%20recognition-by-components%20a%20theory%20of%20human%20image%20understanding.pdf](http://wexler.free.fr/library/files/biederman%20(1987)%20recognition-by-components%20a%20theory%20of%20human%20image%20understanding.pdf)



Recognition is all about modeling variability

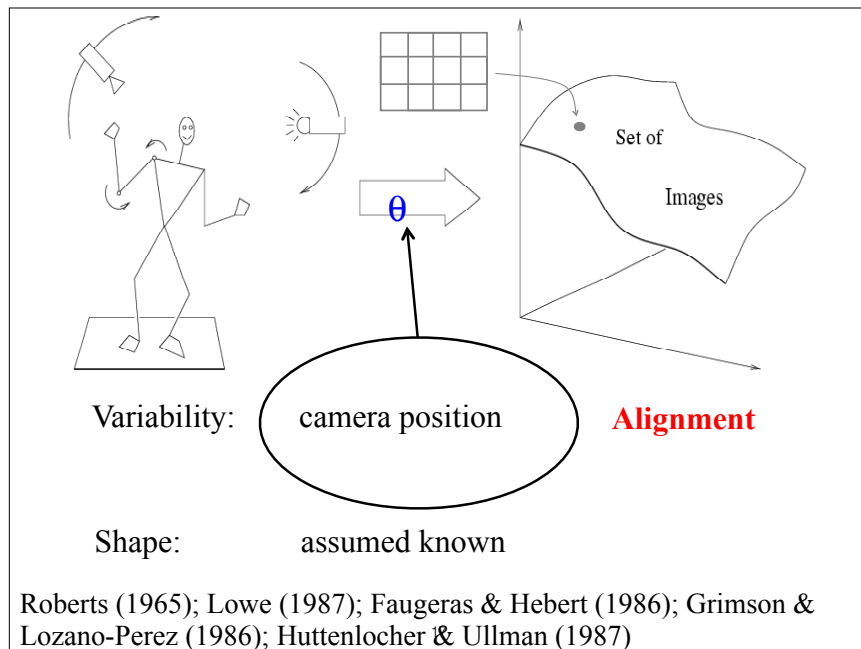


Variability:

- Camera position
- Illumination
- Within-class variation
- Background, occlusion

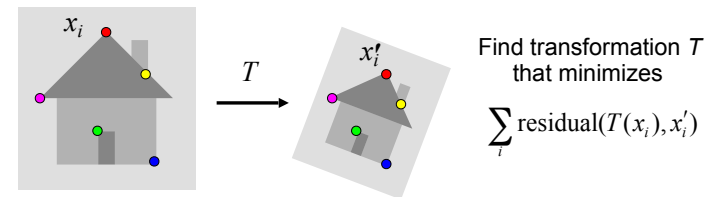
History of ideas in recognition

1960s – early 1990s: the geometric era



Recall: Alignment

Alignment: fitting a model to a transformation between pairs of features (*matches*) in two images

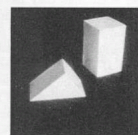


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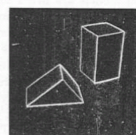
Recognition as an alignment problem: Block world



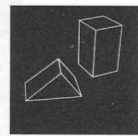
a)



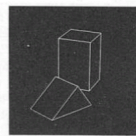
b)



c)



d)

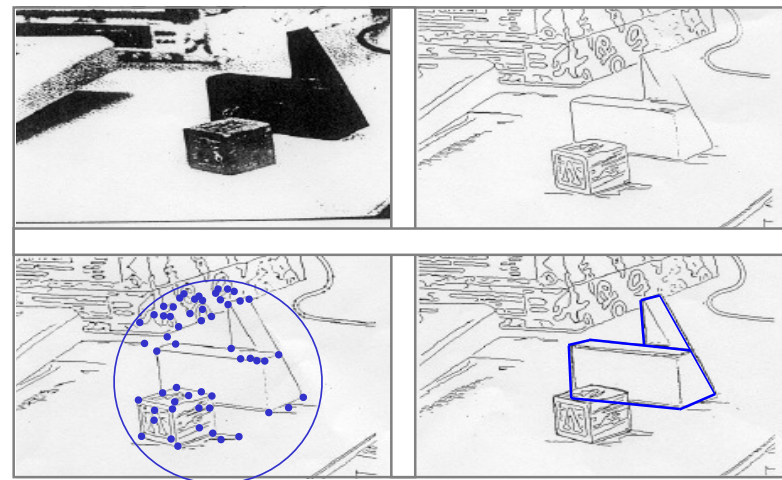


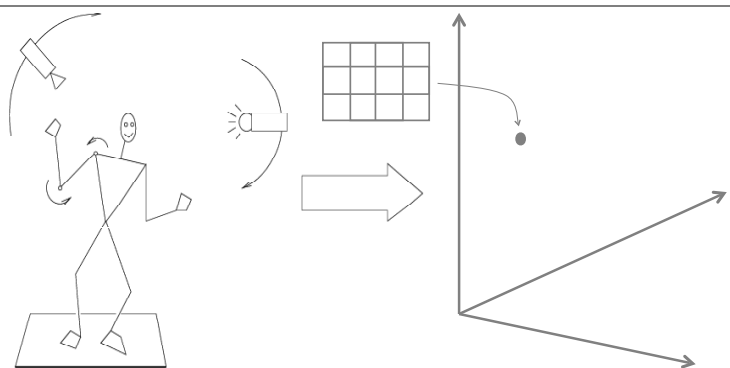
e)

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

Fig. 1. A system for recognizing 3-d polyhedral scenes. a) L.G. Roberts. b) A blocks world scene. c) Detected edges using a 2x2 gradient operator. d) A 3-d polyhedral description of the scene, formed automatically from the single image. e) The 3-d scene displayed with a viewpoint different from the original image to demonstrate its accuracy and completeness. (b) - e) are taken from [64] with permission MIT Press.)

Alignment: Huttenlocher & Ullman (1987)



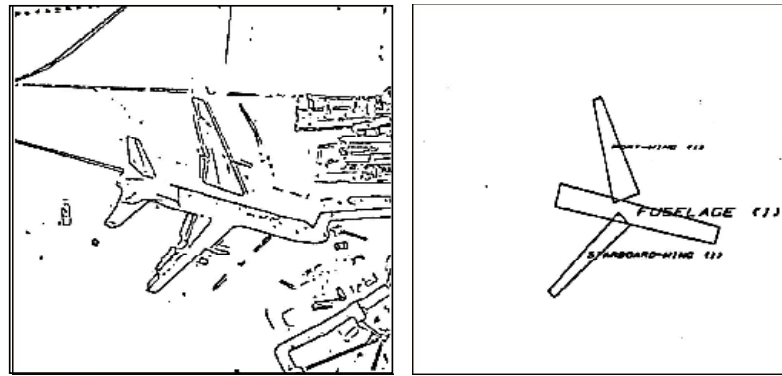


~~Variability~~ **Invariance to:** Camera position
Illumination
Etc.

Duda & Hart (1972); Weiss (1987); Mundy et al. (1992-94);
Rothwell et al. (1992); Burns et al. (1993)

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From object instances to object categories

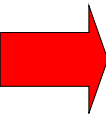
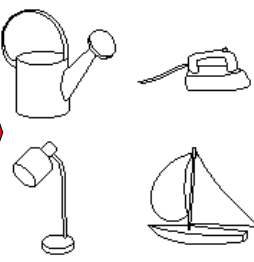


ACRONYM (Brooks and Binford, 1981)
Binford (1971), Nevatia & Binford (1972), Marr & Nishihara (1978)

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Recognition by components

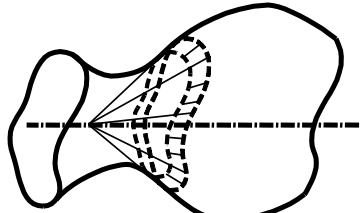
Biederman (1987)

Primitives (geons)					Objects
Cube Straight Edge Straight Axis Constant	Wedge Straight Edge Straight Axis Expanded	Pyramid Straight Edge Straight Axis Expanded	Cylinder Curved Edge Straight Axis Constant	Barrel Curved Edge Straight Axis Exp & Cont	 
Arch Straight Edge Curved Axis Constant	Cone Curved Edge Straight Axis Expanded	Expanded Cylinder Curved Edge Straight Axis Expanded	Handle Curved Edge Curved Axis Constant	Expanded Handle Curved Edge Curved Axis Expanded	

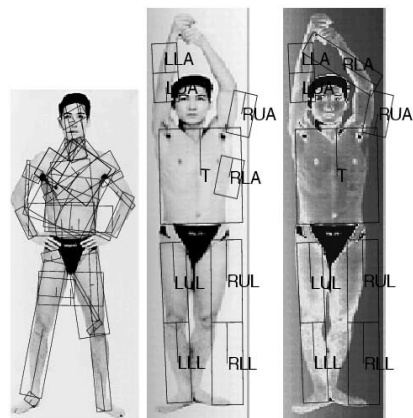
http://en.wikipedia.org/wiki/Recognition_by_Components_Theory

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General shape primitives?



Generalized cylinders
Ponce et al. (1989)



Zisserman et al. (1995)

Forsyth (2000)

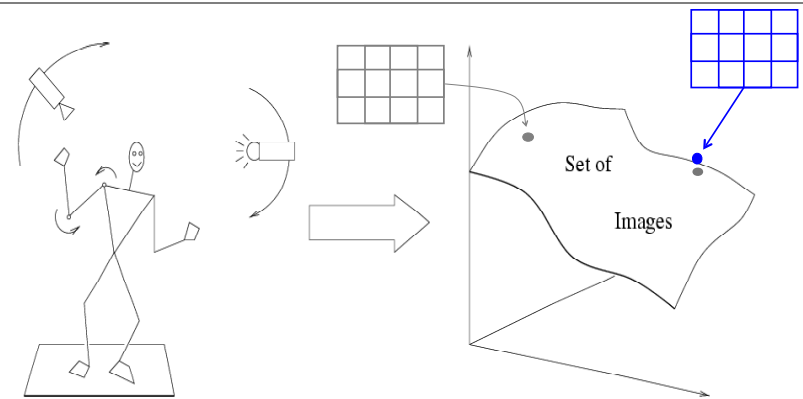
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History of ideas in recognition

1960s – early 1990s: the geometric era

1990s: appearance-based models

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Empirical models of image variability

Appearance-based techniques

Turk & Pentland (1991); Murase & Nayar (1995); etc.

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Eigenfaces (Turk & Pentland, 1991)



Experimental Condition	Correct/Unknown Recognition Percentage		
	Lighting	Orientation	Scale
Forced classification	96/0	85/0	64/0
Forced 100% accuracy	100/19	100/39	100/60
Forced 20% unknown rate	100/20	94/20	74/20

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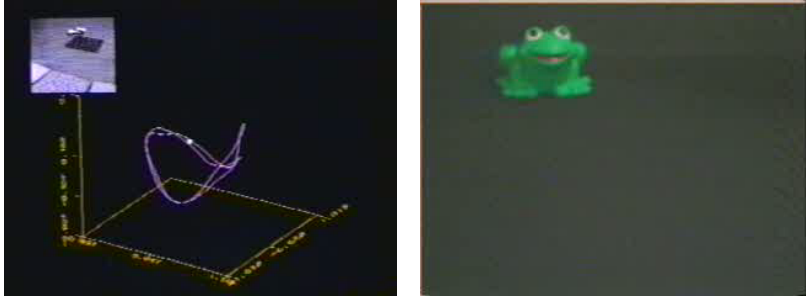
Color Histograms



Swain and Ballard, [Color Indexing](#), IJCV 1991.

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Appearance manifolds



H. Murase and S. Nayar, Visual learning and recognition of 3-d objects from appearance, IJCV 1995

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Limitations of global appearance models

Requires global registration of patterns

Not robust to clutter, occlusion, geometric transformations



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History of ideas in recognition

1960s – early 1990s: the geometric era

1990s: appearance-based models

1990s – present: sliding window approaches

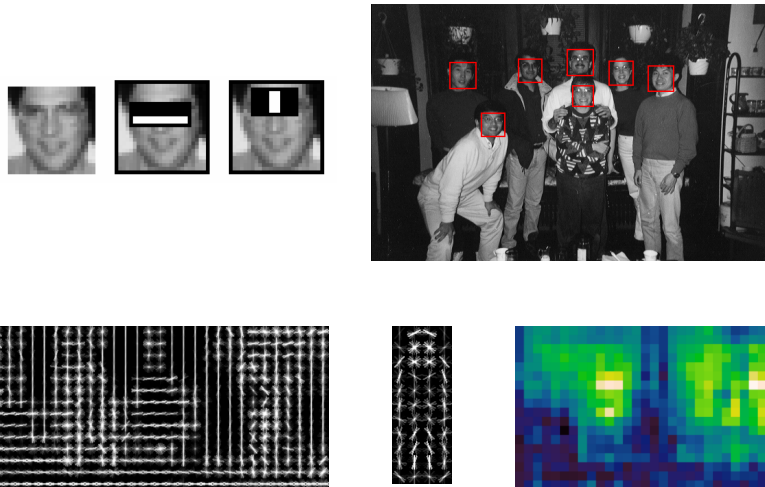
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Sliding window approaches



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Sliding window approaches



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Late 1990s: local features

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Local features for object instance recognition

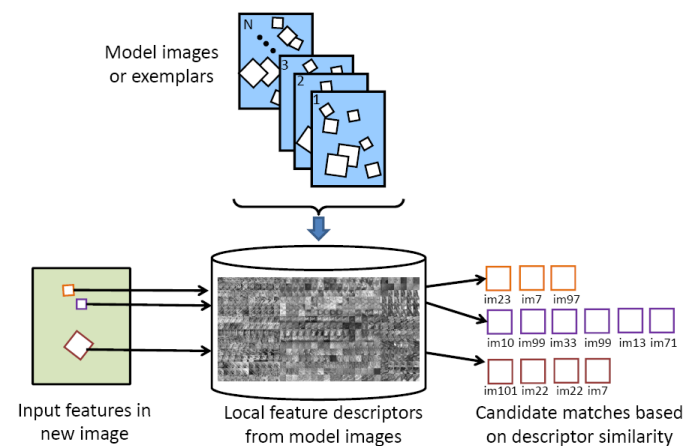


D. Lowe (1999, 2004)

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Large-scale image search

Combining local features, indexing, and spatial constraints

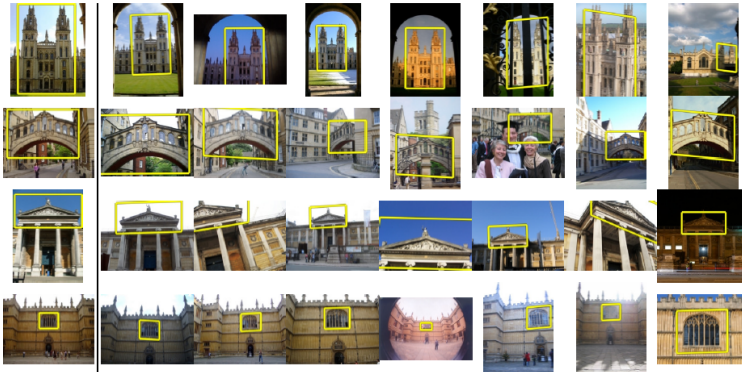


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Image credit: K. Grauman and B. Leibe

Large-scale image search

Combining local features, indexing, and spatial constraints



Philbin et al. '07

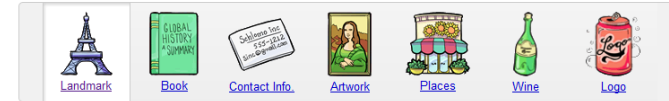
37

Large-scale image search

Combining local features, indexing, and spatial constraints

Google Goggles in Action

Click the icons below to see the different ways Google Goggles can be used.



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Late 1990s: local features

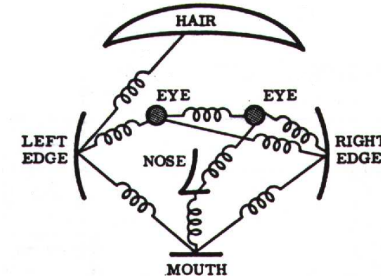
Early 2000s: parts-and-shape models

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Parts-and-shape models

Model:

- Object as a set of parts
- Relative locations between parts
- Appearance of part



Fischler & Eischlager 73 40

Constellation models

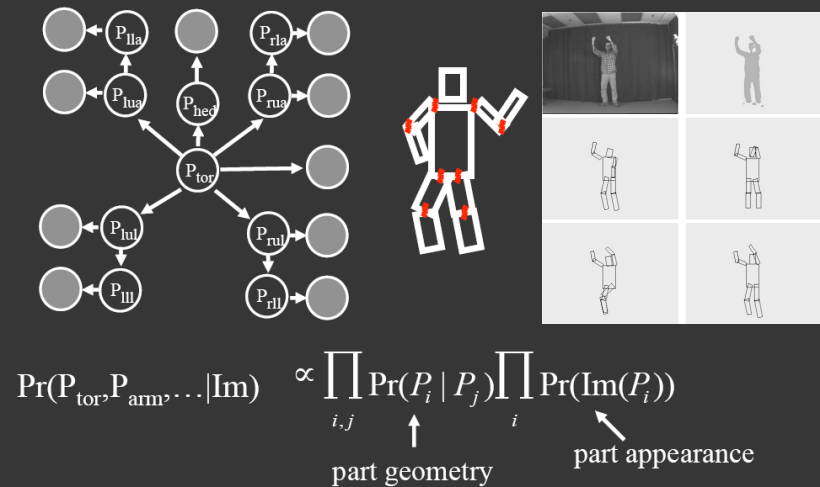


Weber, Welling & Perona (2000), Fergus, Perona & Zisserman (2003)

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Pictorial structure model

Fischler and Elschlager(73), Felzenszwalb and Huttenlocher(00)



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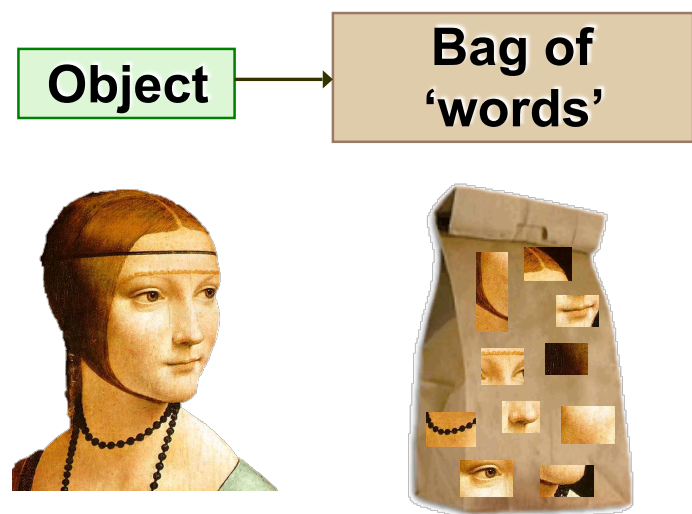
Late 1990s: local features

Early 2000s: parts-and-shape models

Mid/Late-2000s: bags of features, fully learned models

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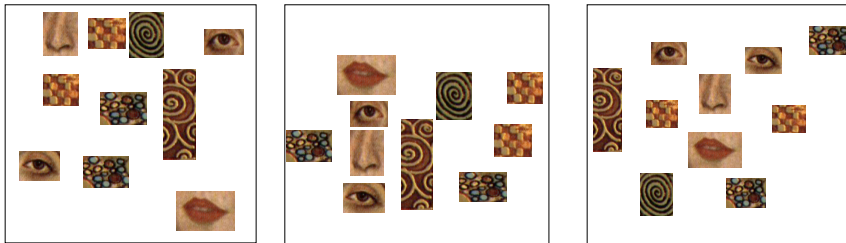
Bag-of-features models



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Objects as texture

All of these are treated as being the same



No distinction between foreground and background: scene recognition?

Learning algorithms to the rescue.

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Learned part-based models



Poselet detectors: Bourdev, Maji and Malik



Deformable part-based models, Girshick, Felzenszwalb, Ramanan, McAllester

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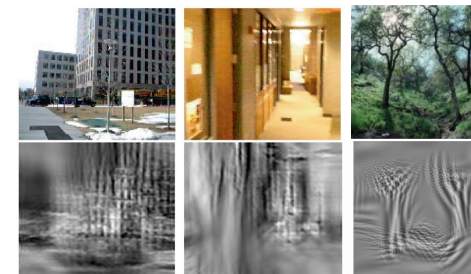
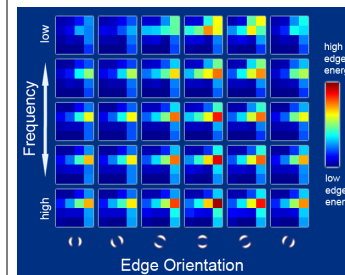
Mid-2000s: bags of features

Present trends: “big data”, context, attributes, combining geometry and recognition, advanced scene understanding tasks, deep learning

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Global appearance models revisited

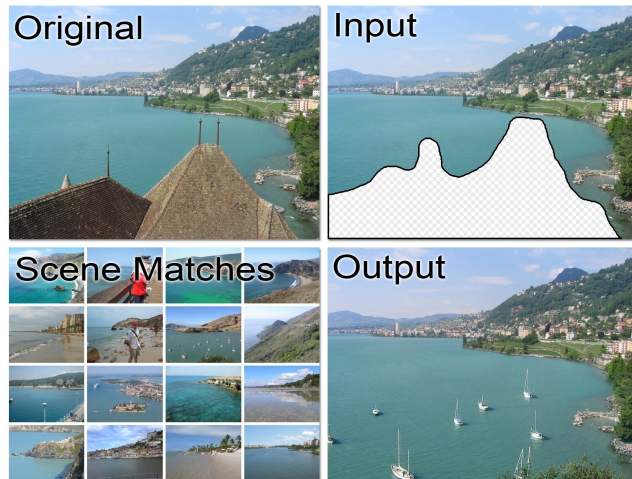
The “gist” of a scene: Oliva & Torralba (2001)



<http://people.csail.mit.edu/torralba/code/spatialenvelope/>

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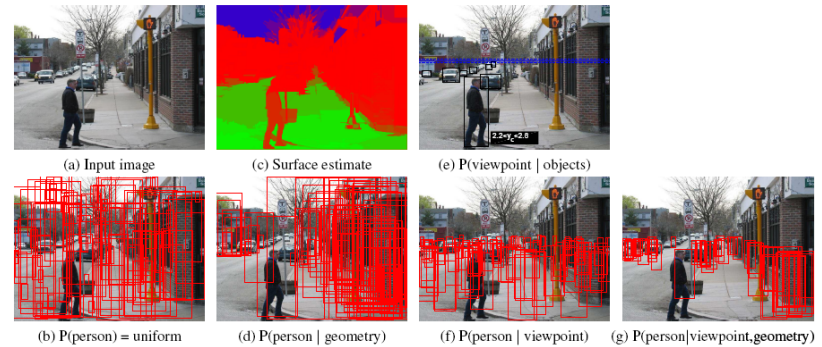
New applications in graphics



J. Hays and A. Efros, [Scene Completion using Millions of Photographs](#), SIGGRAPH 2007

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Geometric context



D. Hoiem, A. Efros, and M. Herbert, [Putting Objects in Perspective](#), CVPR 2006

50

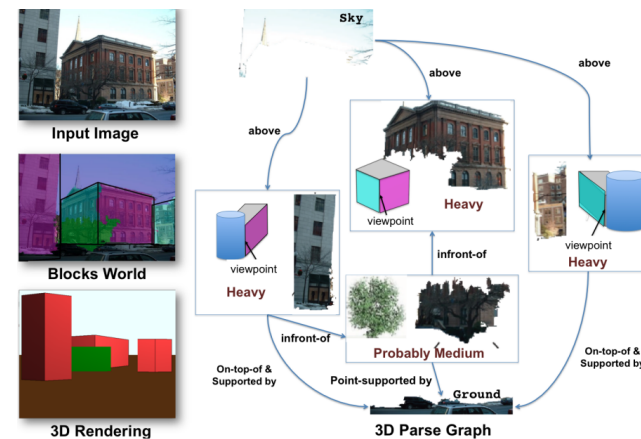
Geometry and recognition



V. Hedau, D. Hoiem, and D. Forsyth, [Recovering the Spatial Layout of Cluttered Rooms](#), ICCV 2009.

51

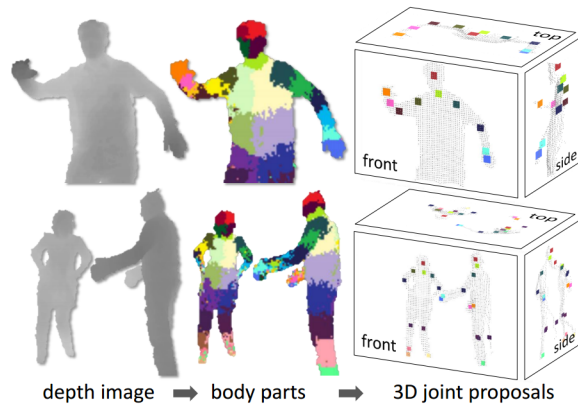
Geometry and recognition



A. Gupta, A. Efros and M. Hebert, [Blocks World Revisited: Image Understanding Using Qualitative Geometry and Mechanics](#), ECCV 2010

52

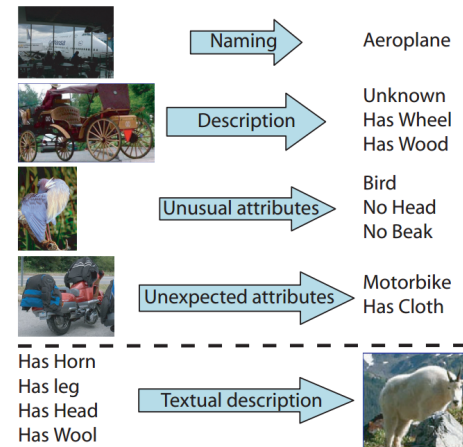
Recognition from RGBD Images



J. Shotton, A. Fitzgibbon, M. Cook, T. Sharp, M. Finocchio, R. Moore, A. Kipman, and A. Blake, [Real-Time Human Pose Recognition in Parts from a Single Depth Image](#), CVPR 2011

53

Attributes for recognition



A. Farhadi, I. Endres, D. Hoiem, and D Forsyth, [Describing Objects by their Attributes](#), CVPR 2009

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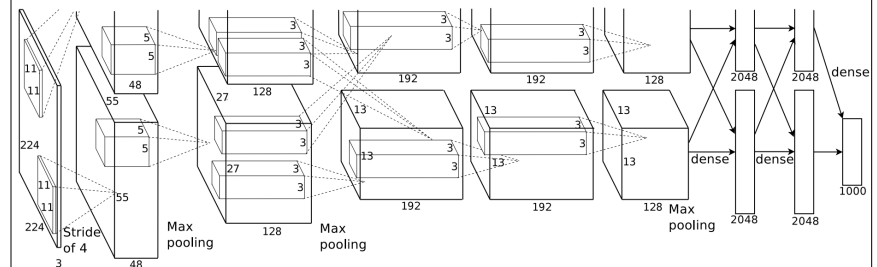
Deep learning



[NY Times article](#)

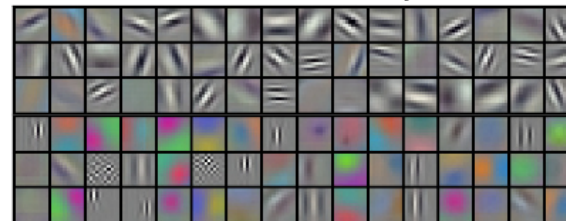
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Recent deep learning breakthroughs...



[ImageNet Classification with Deep Convolutional Neural Networks](#) Alex Krizhevsky, Ilya Sutskever, Geoffrey E. Hinton NIPS 2014

96 filters learned in layer 1



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Further thoughts and readings

- Chapter 14, Szeliski's book
- Think of the applications of computer vision around you