

Course Administration

- Project 2 results are online
- Project 3 is out today
- The first quiz is a week from today (don't panic!)
 - Covers all material up to the quiz
 - Emphasizes lecture material NOT project topics
 - No calculators needed; no formulas need to be memorized; testing conceptual understanding mostly
 - Recommended study regime: Attend every lecture, go over the lecture slides once, go over the textbook sections once.



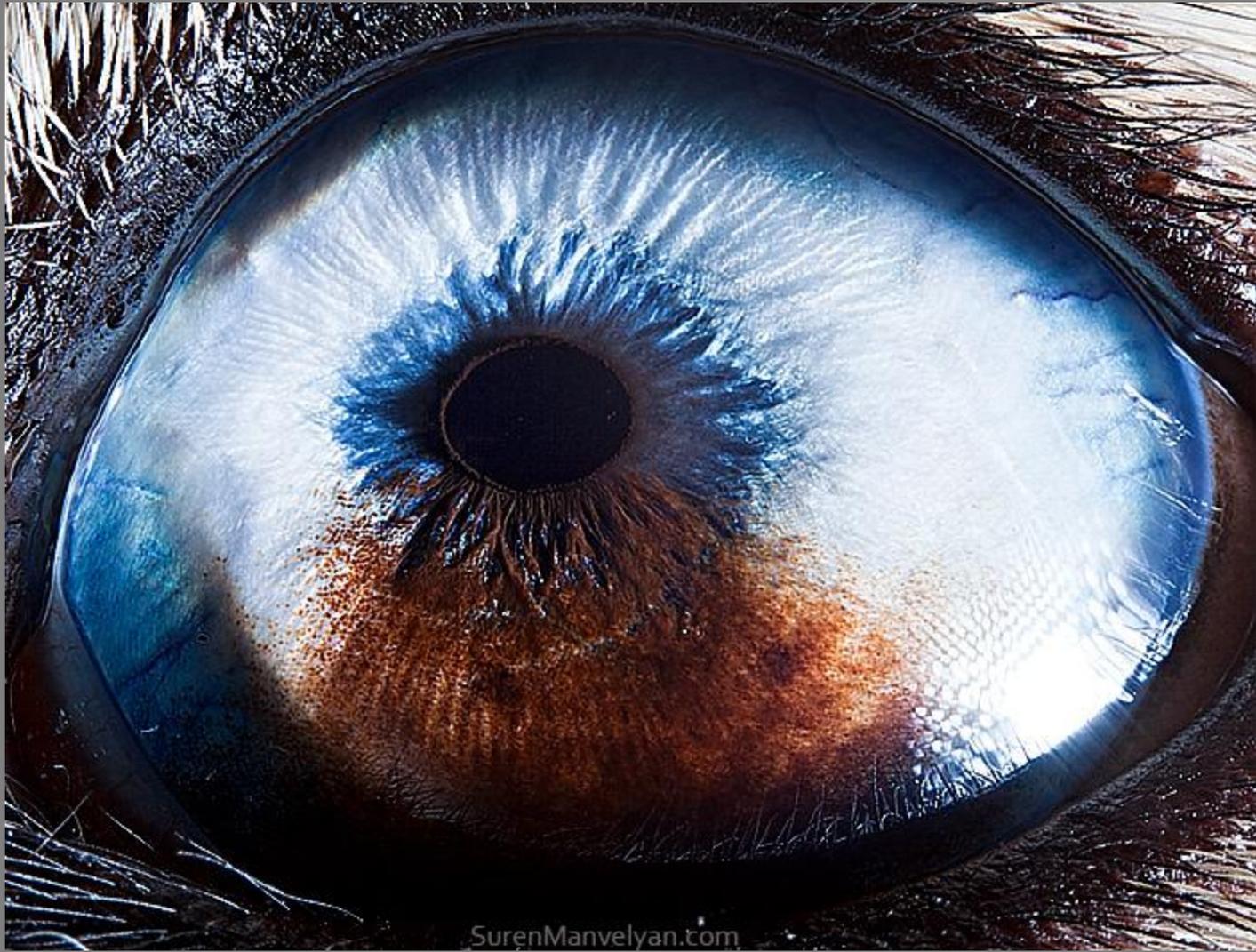
Suren Manvelyan



Suren Manvelyan



Suren Manvelyan



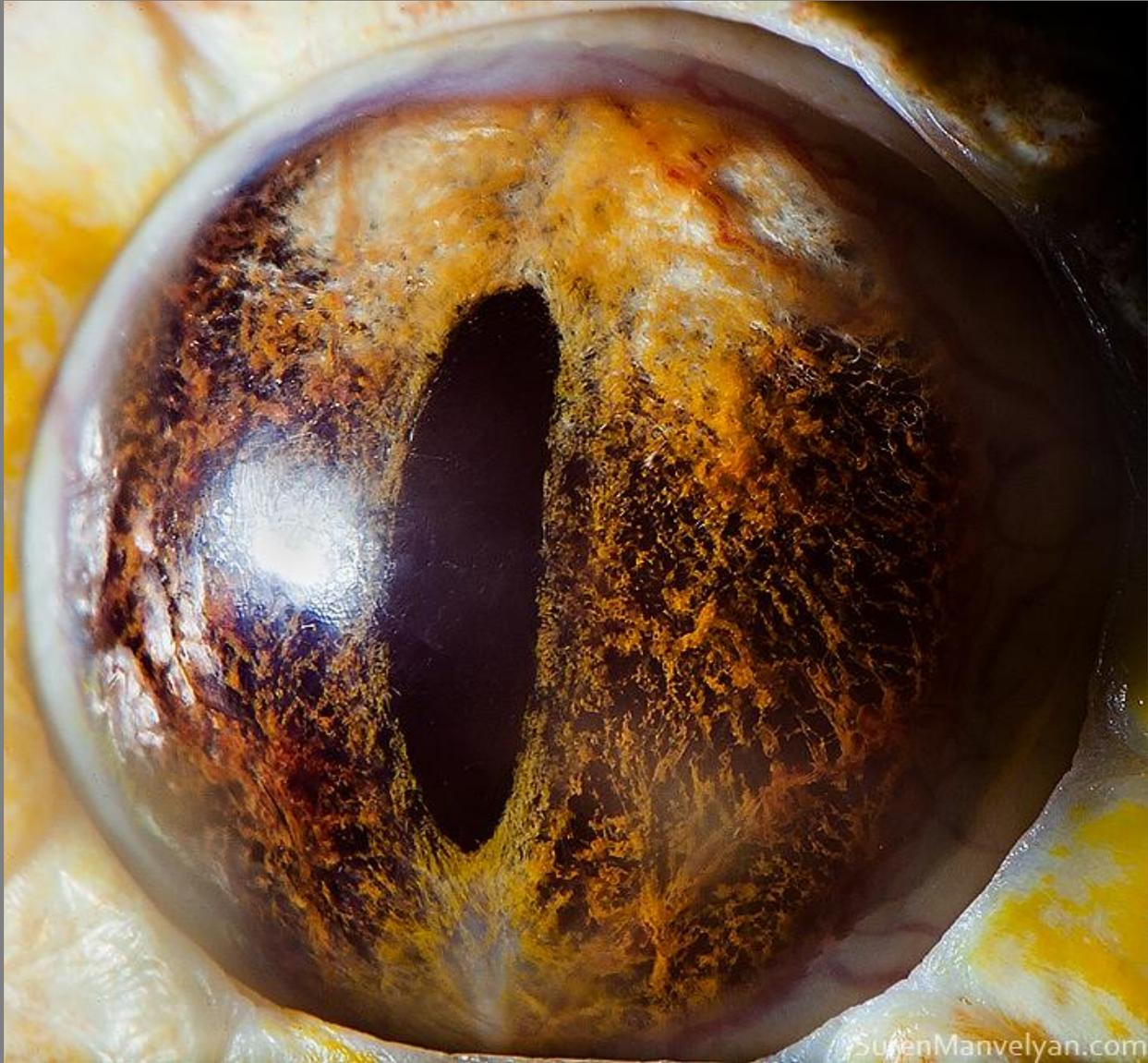
SurenManvelyan.com







By Suren Manvelyan, <http://www.surenmanvelyan.com/gallery/7116>



SurenManvelyan.com



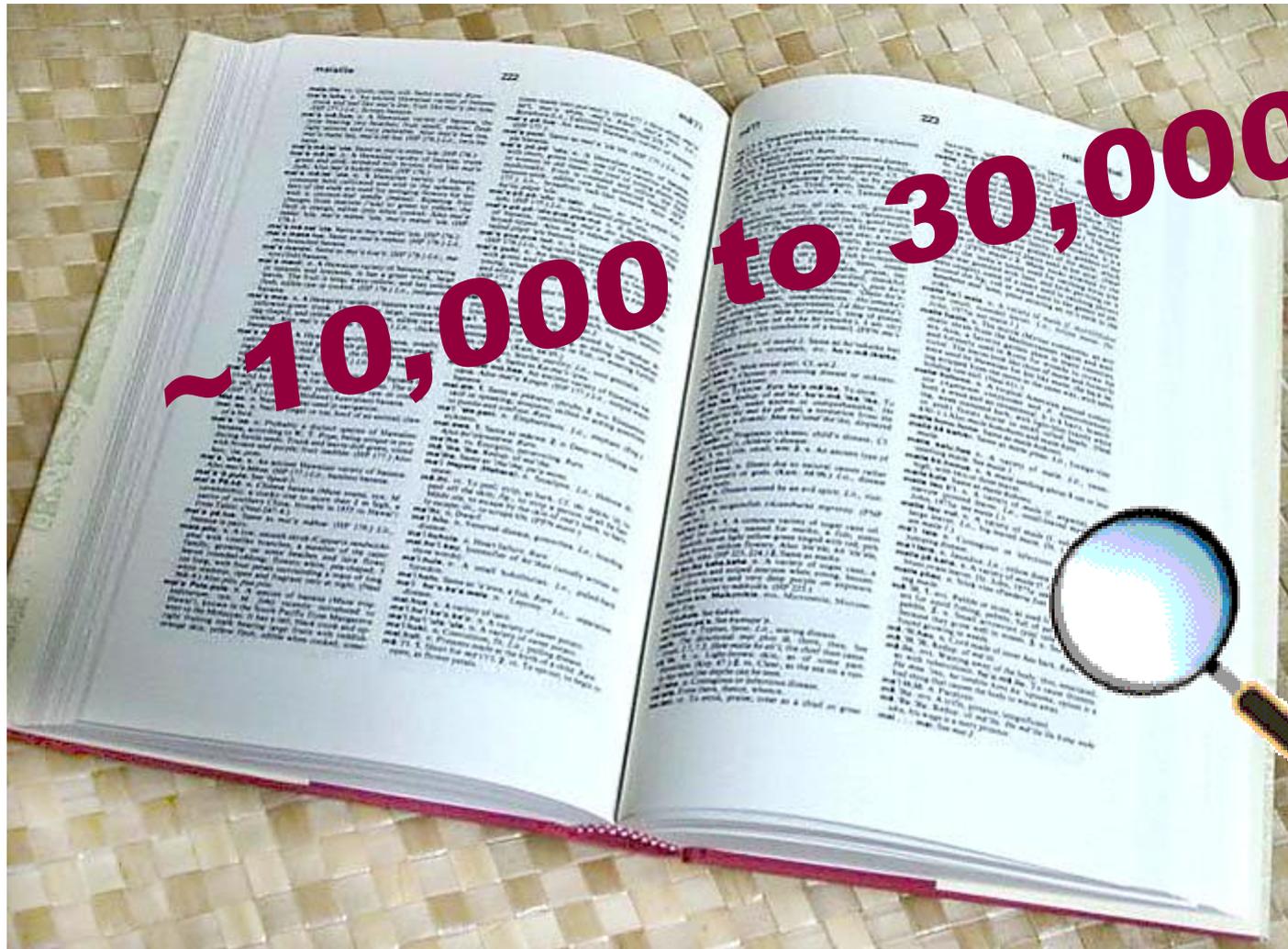
Suren Manvelyan

Recognition: Overview and History



Slides from Lana Lazebnik, Fei-Fei Li, Rob Fergus, Antonio Torralba, and Jean Ponce

How many visual object categories are there?

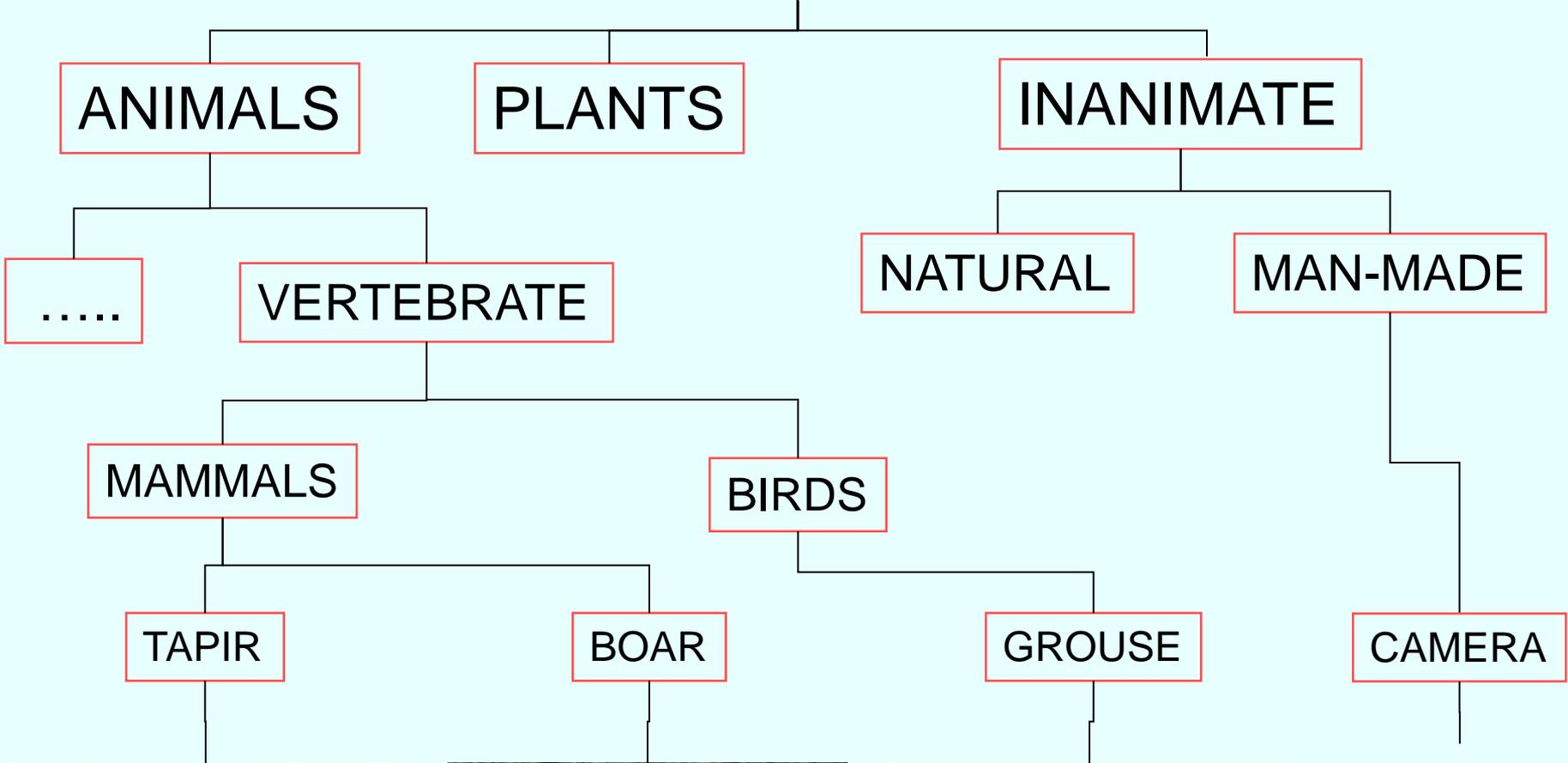




~10,000 to 30,000



OBJECTS



Specific recognition tasks



Scene categorization or classification

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



Image annotation / tagging / attributes



- street
- people
- building
- mountain
- tourism
- cloudy
- brick
- ...

Object detection

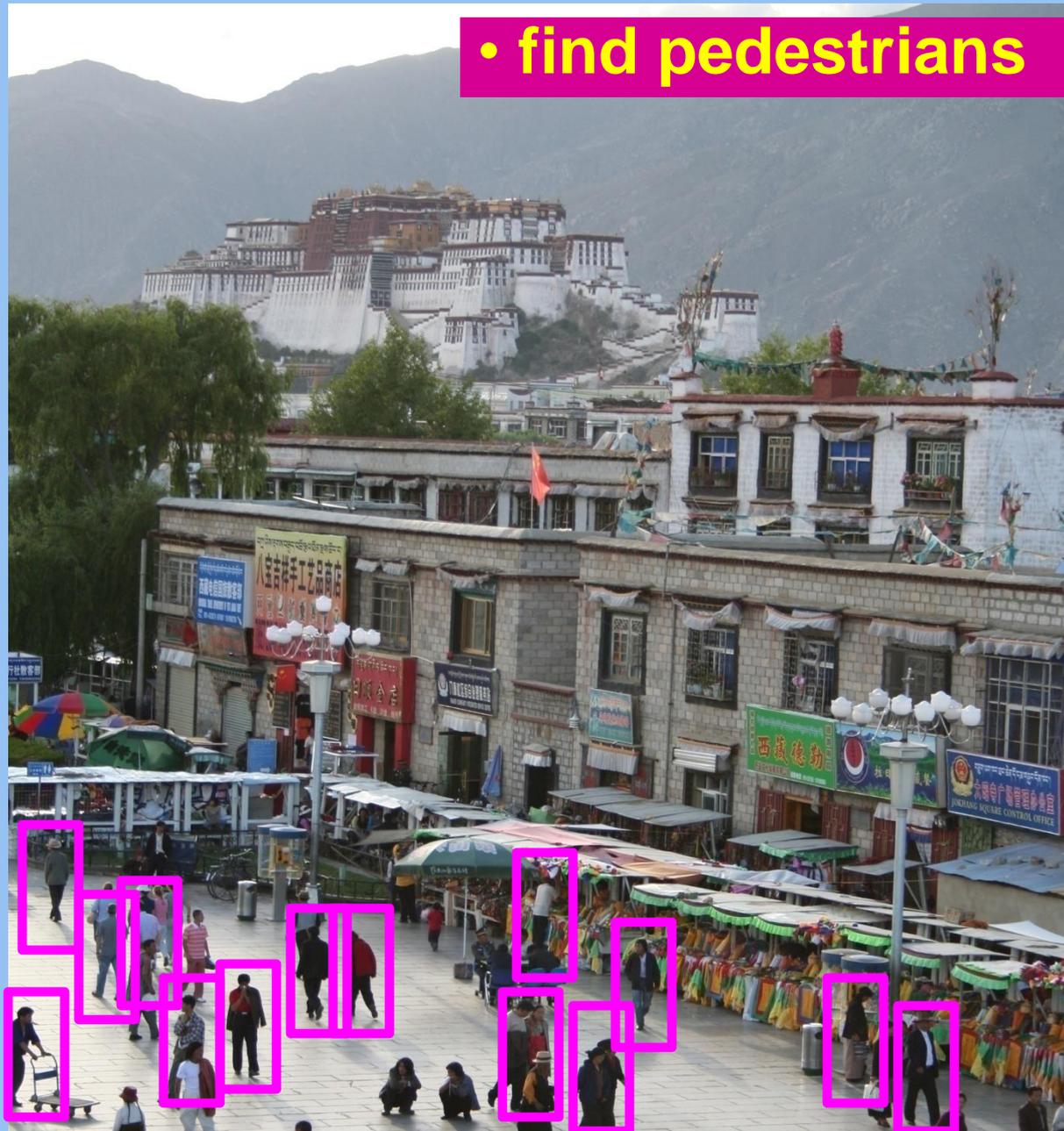


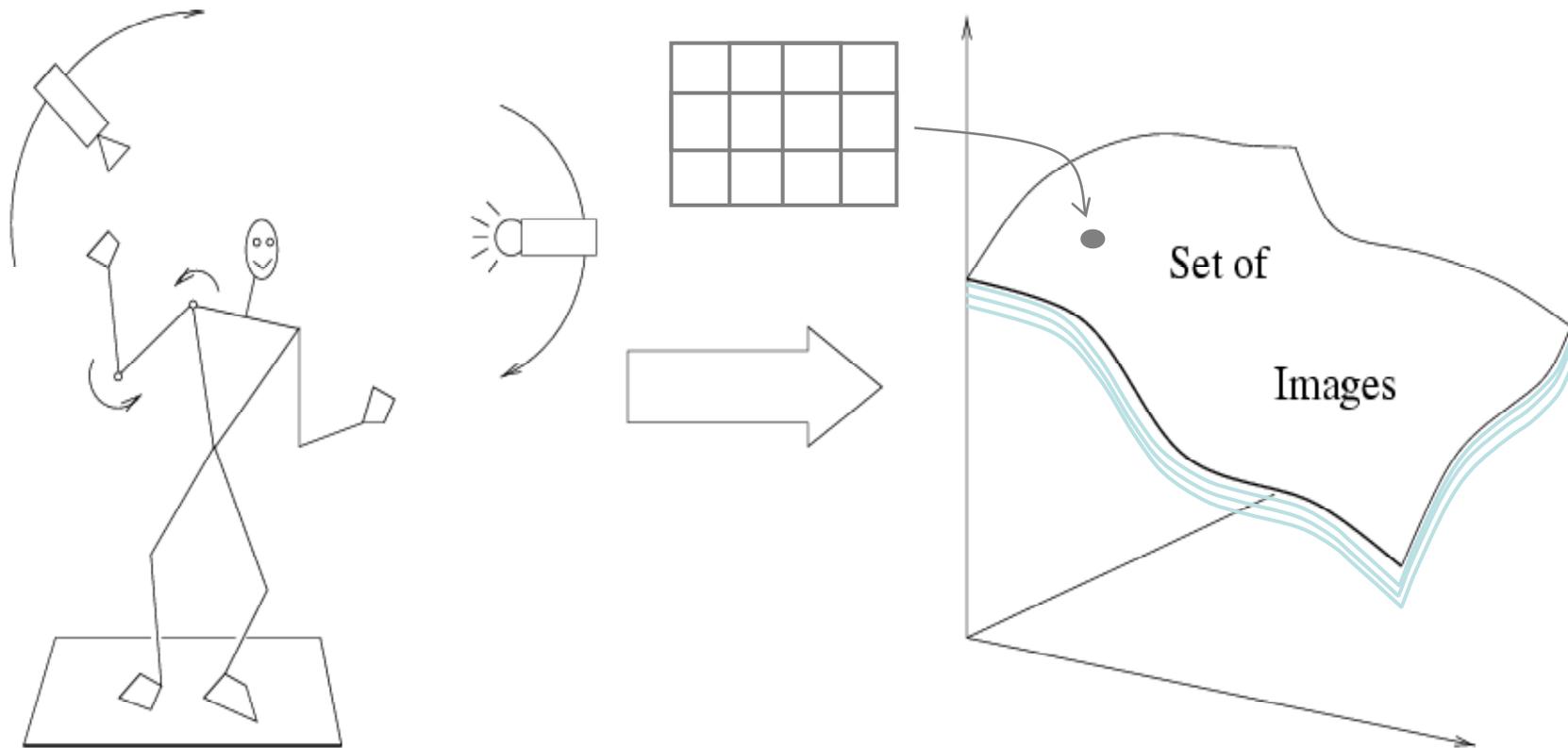
Image parsing



Scene understanding?



Recognition is all about modeling variability

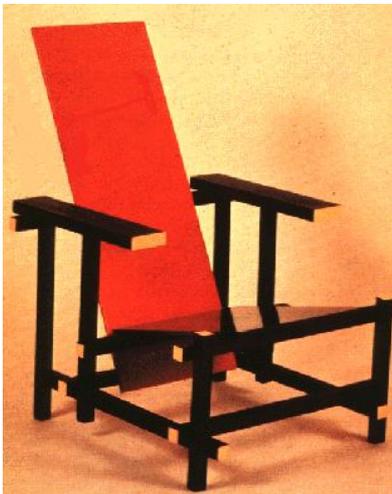


Variability: Camera position
Illumination
Shape parameters



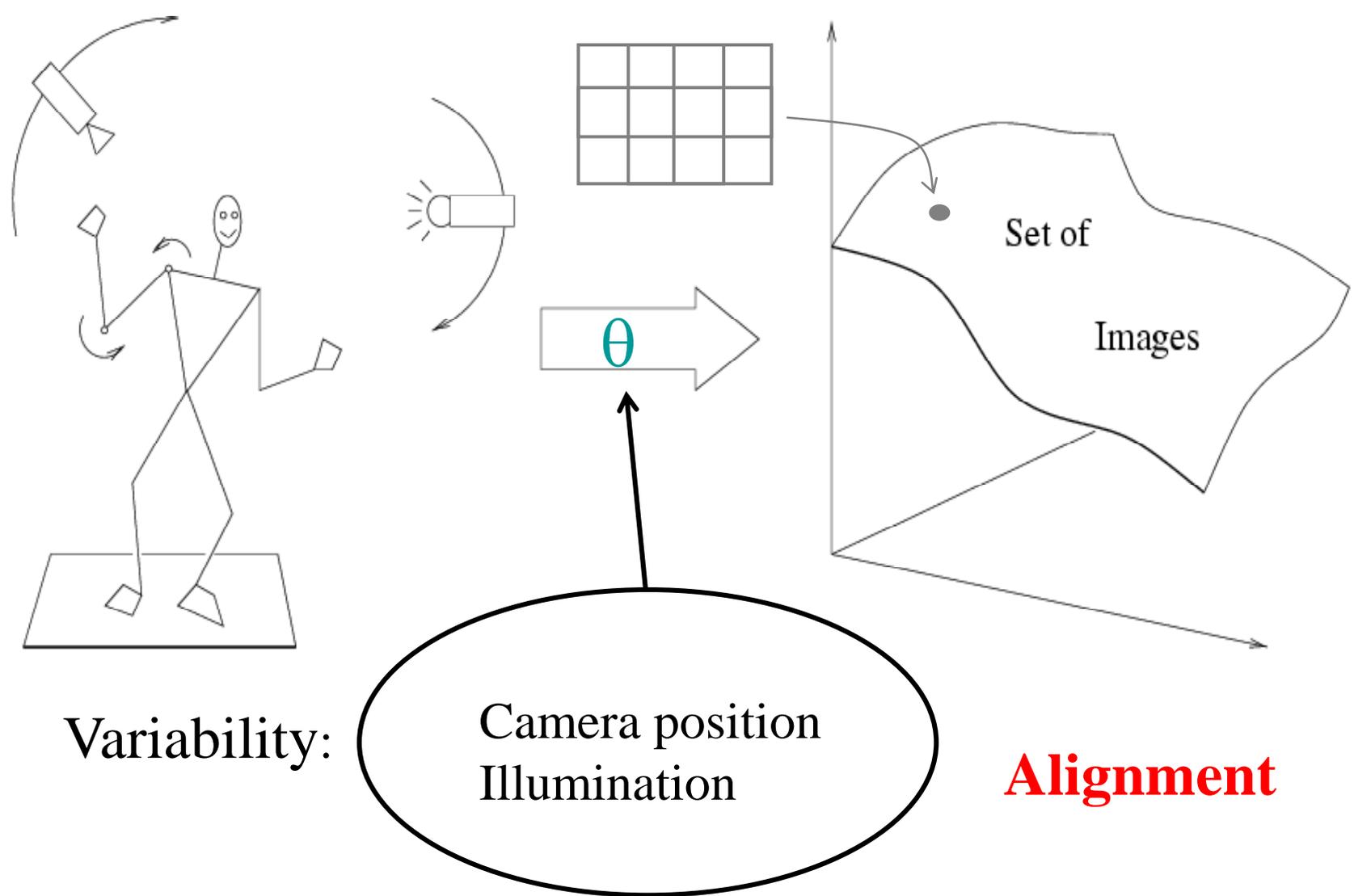
Within-class variations?

Within-class variations



History of ideas in recognition

- 1960s – early 1990s: the geometric era

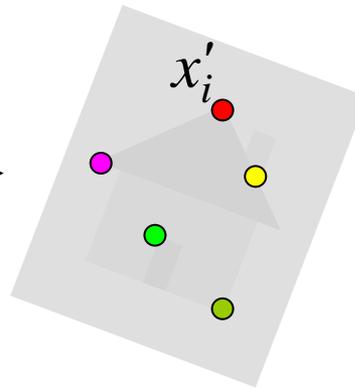
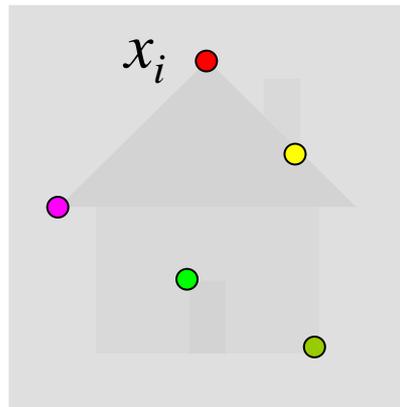


Shape: assumed known

Roberts (1965); Lowe (1987); Faugeras & Hebert (1986); Grimson & Lozano-Perez (1986);
Huttenlocher & Ullman (1987)

Recall: Alignment

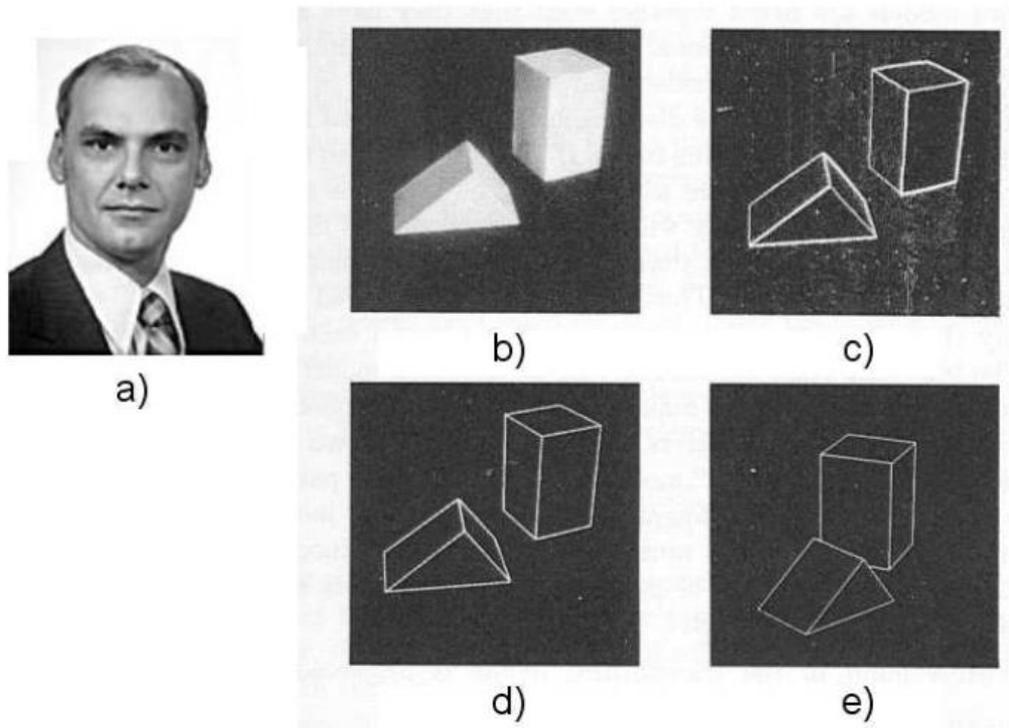
- Alignment: fitting a model to a transformation between pairs of features (*matches*) in two images



Find transformation T
that minimizes

$$\sum_i \text{residual}(T(x_i), x'_i)$$

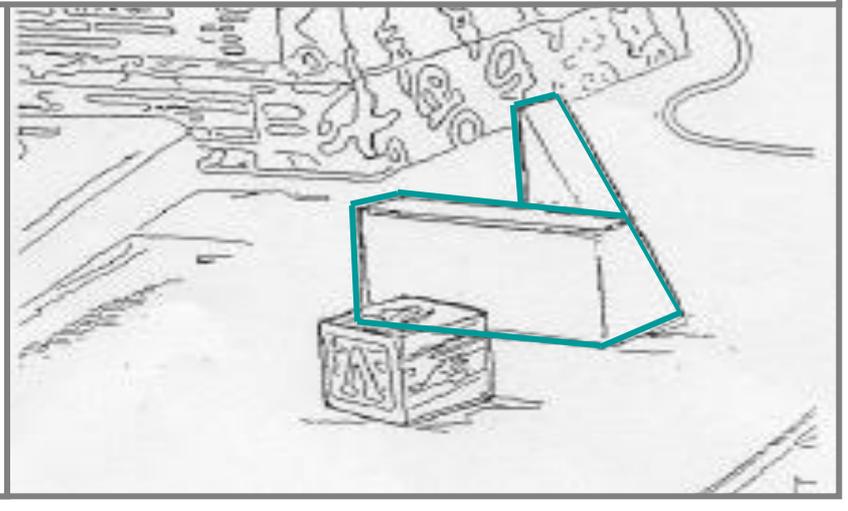
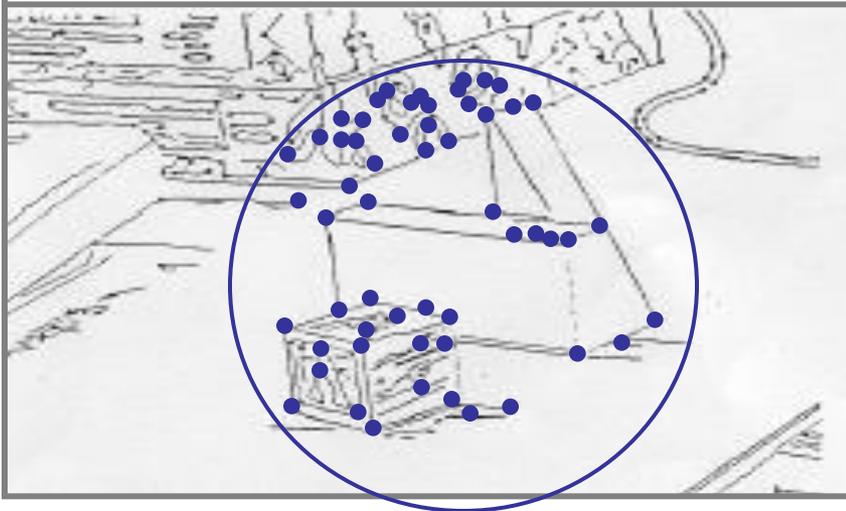
Recognition as an alignment problem: Block world



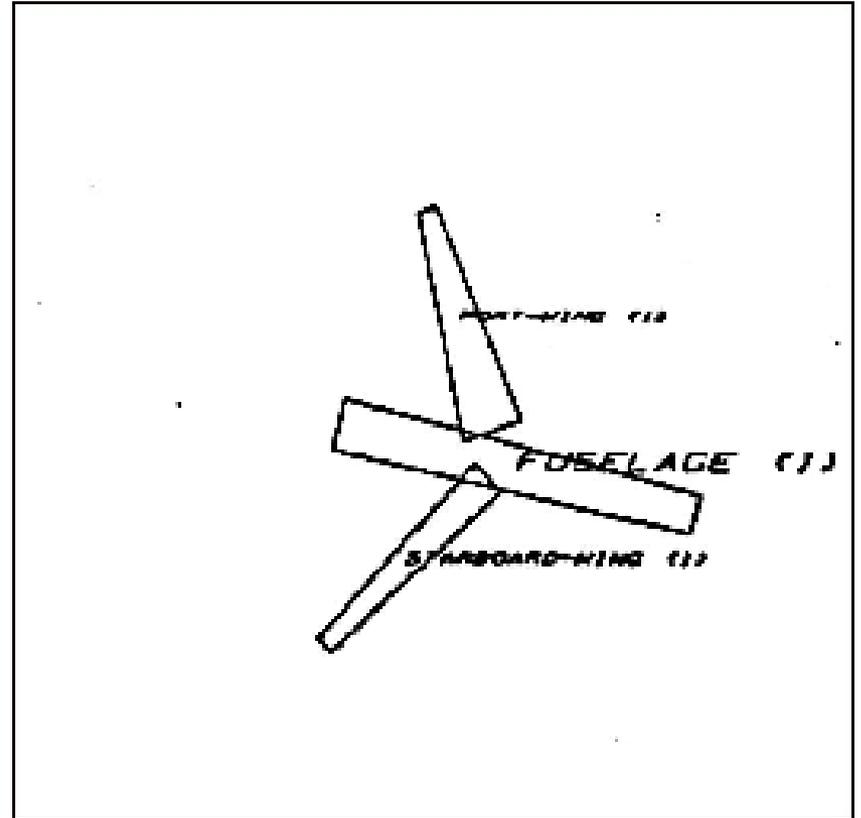
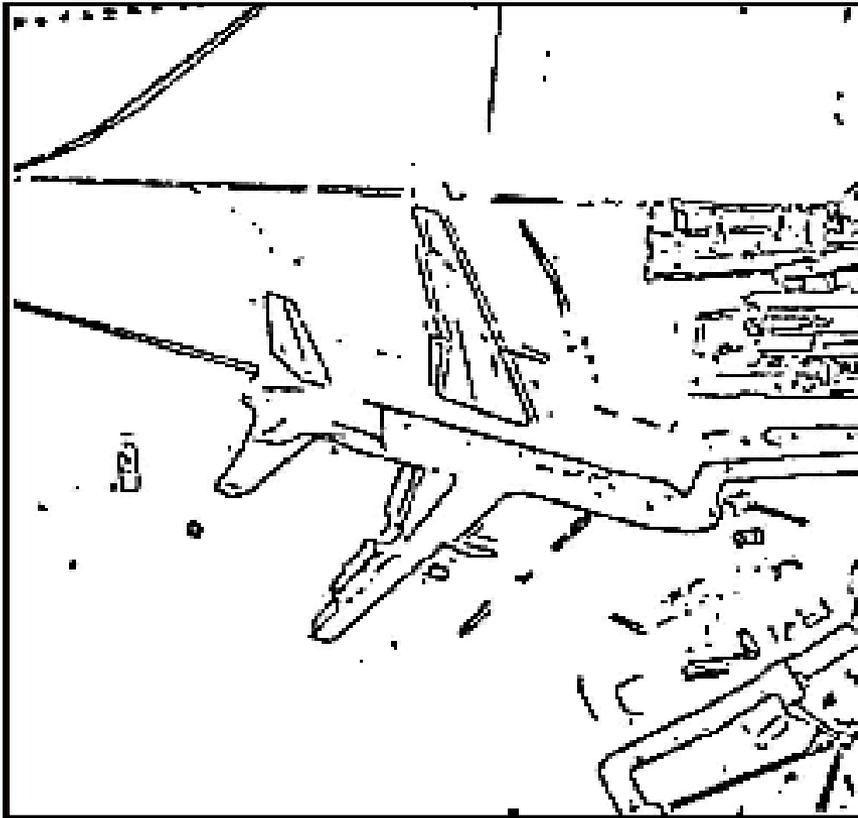
L. G. Roberts, [*Machine Perception of Three Dimensional Solids*](#), Ph.D. thesis, MIT Department of Electrical Engineering, 1963.

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)



Representing and recognizing object categories is harder...



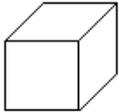
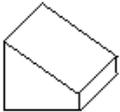
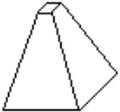
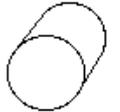
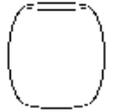
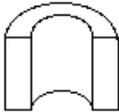
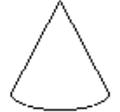
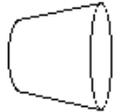
ACRONYM (Brooks and Binford, 1981)

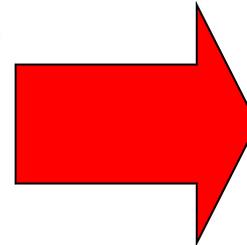
Binford (1971), Nevatia & Binford (1972), Marr & Nishihara (1978)

Recognition by components

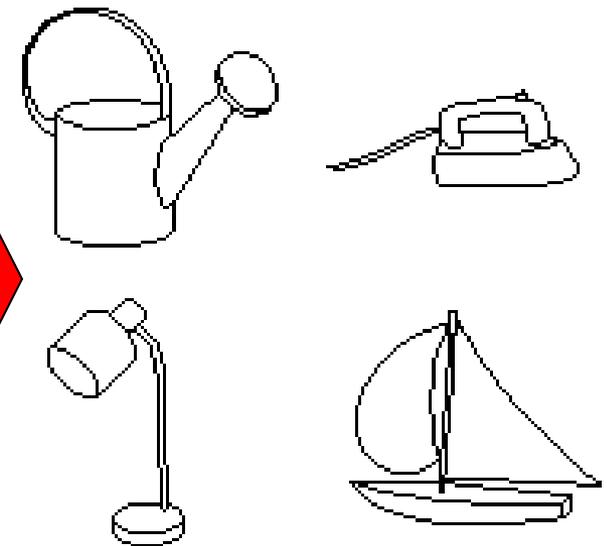
Biederman (1987)

Primitives (geons)

<p>Cube</p>  <p>Straight Edge Straight Axis Constant</p>	<p>Wedge</p>  <p>Straight Edge Straight Axis Expanded</p>	<p>Pyramid</p>  <p>Straight Edge Straight Axis Expanded</p>	<p>Cylinder</p>  <p>Curved Edge Straight Axis Constant</p>	<p>Barrel</p>  <p>Curved Edge Straight Axis Exp & Cont</p>
<p>Arch</p>  <p>Straight Edge Curved Axis Constant</p>	<p>Cone</p>  <p>Curved Edge Straight Axis Expanded</p>	<p>Expanded Cylinder</p>  <p>Curved Edge Straight Axis Expanded</p>	<p>Handle</p>  <p>Curved Edge Curved Axis Constant</p>	<p>Expanded Handle</p>  <p>Curved Edge Curved Axis Expanded</p>

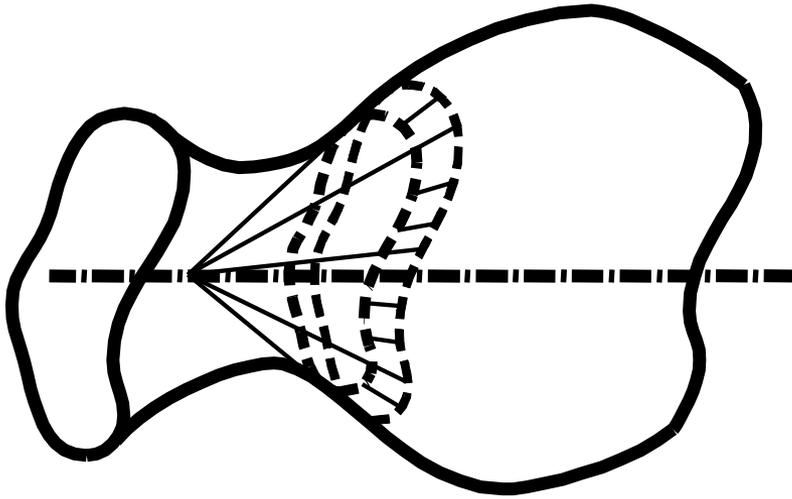


Objects

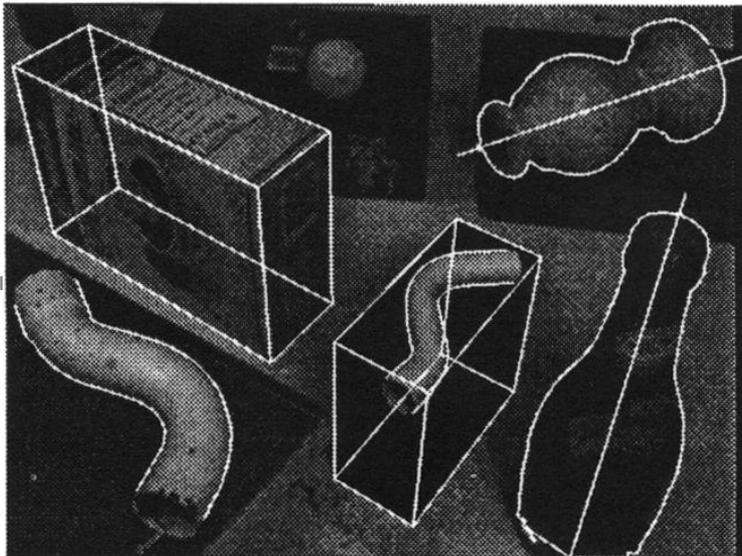


http://en.wikipedia.org/wiki/Recognition_by_Components_Theory

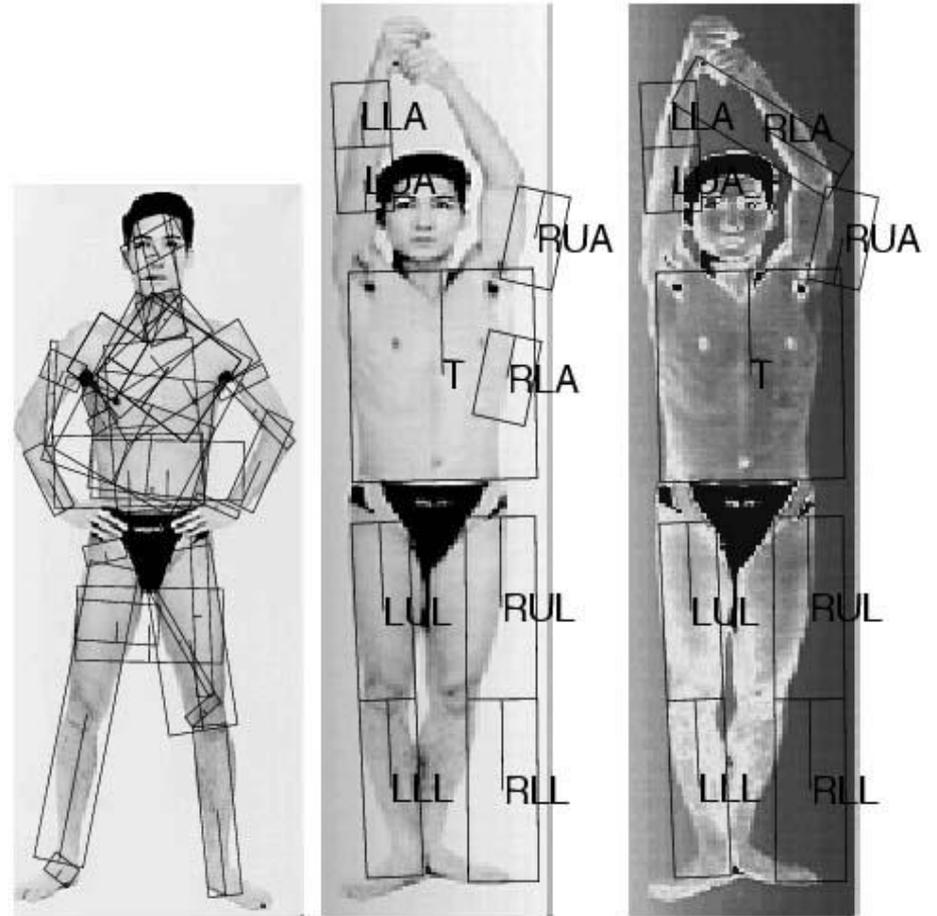
General shape primitives?



Generalized cylinders
Ponce et al. (1989)



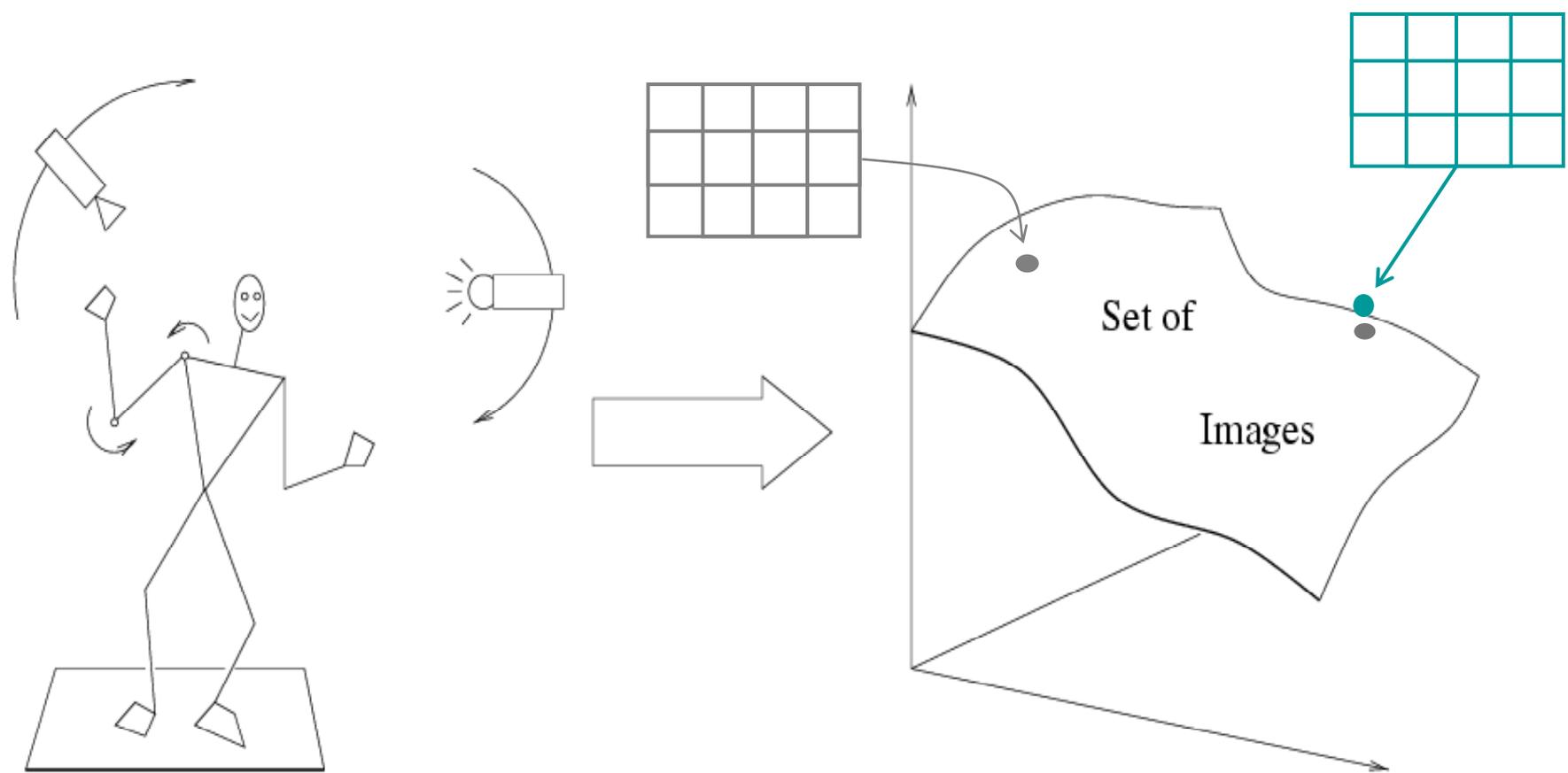
Zisserman et al. (1995)



Forsyth (2000)

History of ideas in recognition

- 1960s – early 1990s: the geometric era
- 1990s: appearance-based models



Empirical models of image variability

Appearance-based techniques

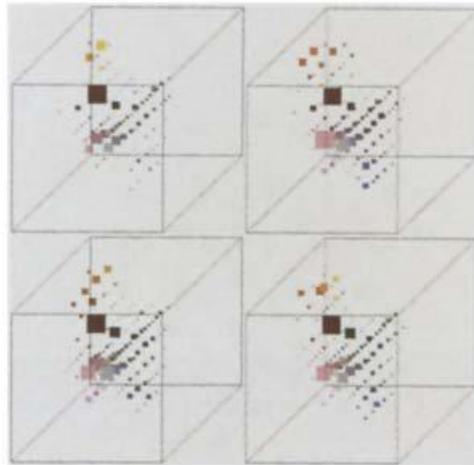
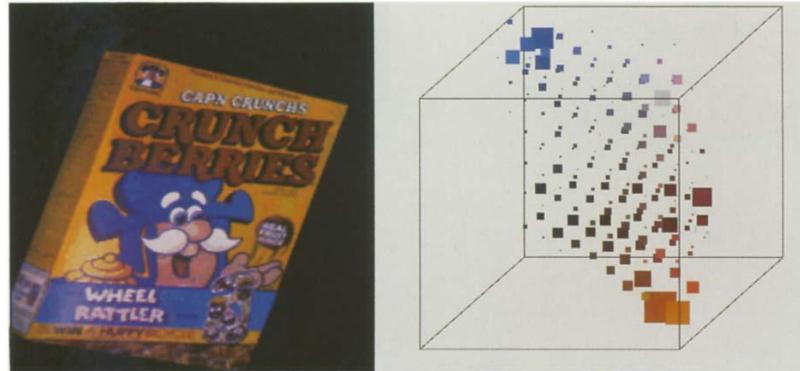
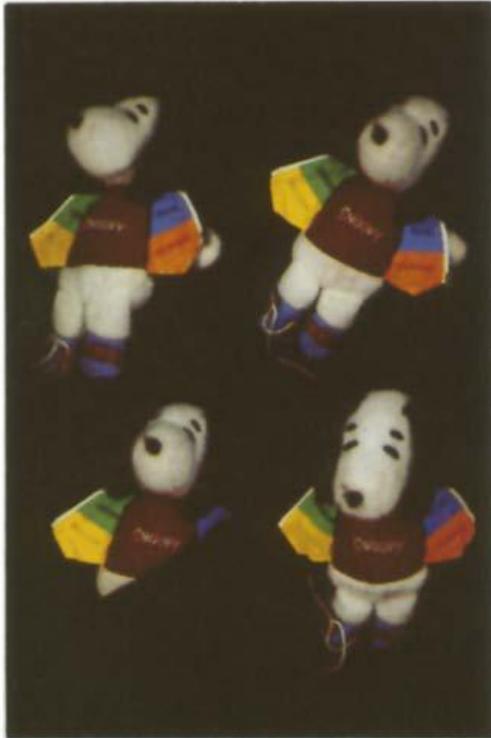
Turk & Pentland (1991); Murase & Nayar (1995); etc.

Eigenfaces (Turk & Pentland, 1991)



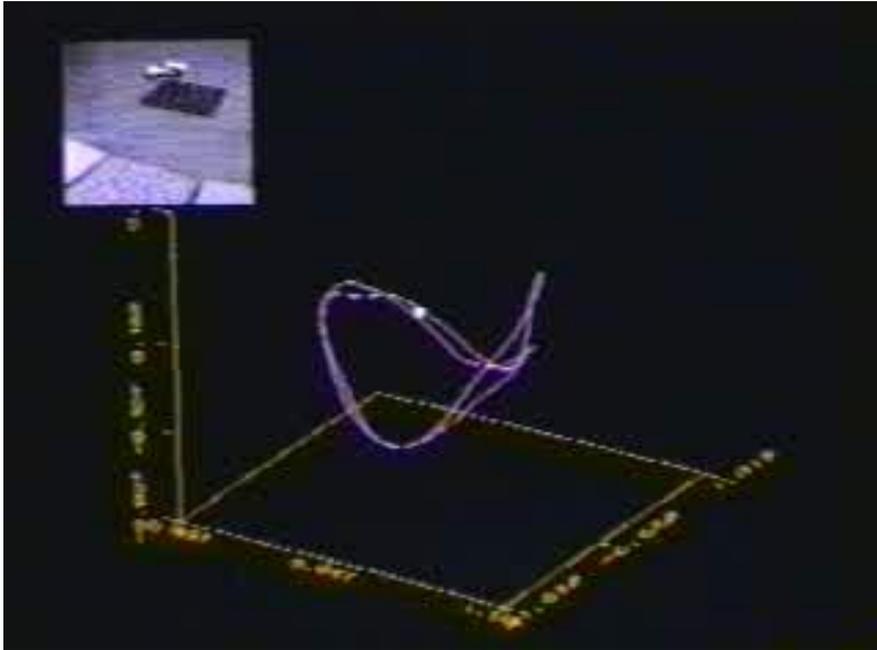
Experimental Condition	Correct/Unknown Recognition Percentage		
	Lighting	Orientation	Scale
Forced classification	96/0	85/0	64/0
Forced 100% accuracy	100/19	100/39	100/60
Forced 20% unknown rate	100/20	94/20	74/20

Color Histograms



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

Appearance manifolds



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

Limitations of global appearance models

- Requires global registration of patterns
- Not robust to clutter, occlusion, geometric transformations



History of ideas in recognition

- 1960s – early 1990s: the geometric era
- 1990s: appearance-based models
- 1990s – present: sliding window approaches

Sliding window approaches



Sliding window approaches



- Turk and Pentland, 1991
- Belhumeur, Hespanha, & Kriegman, 1997
- Schneiderman & Kanade 2004
- Viola and Jones, 2000

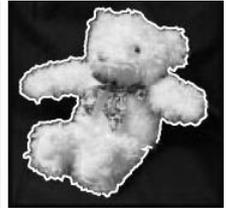


- Schneiderman & Kanade, 2004
- Argawal and Roth, 2002
- Poggio et al. 1993

History of ideas in recognition

- 1960s – early 1990s: the geometric era
- 1990s: appearance-based models
- Mid-1990s: sliding window approaches
- Late 1990s: local features

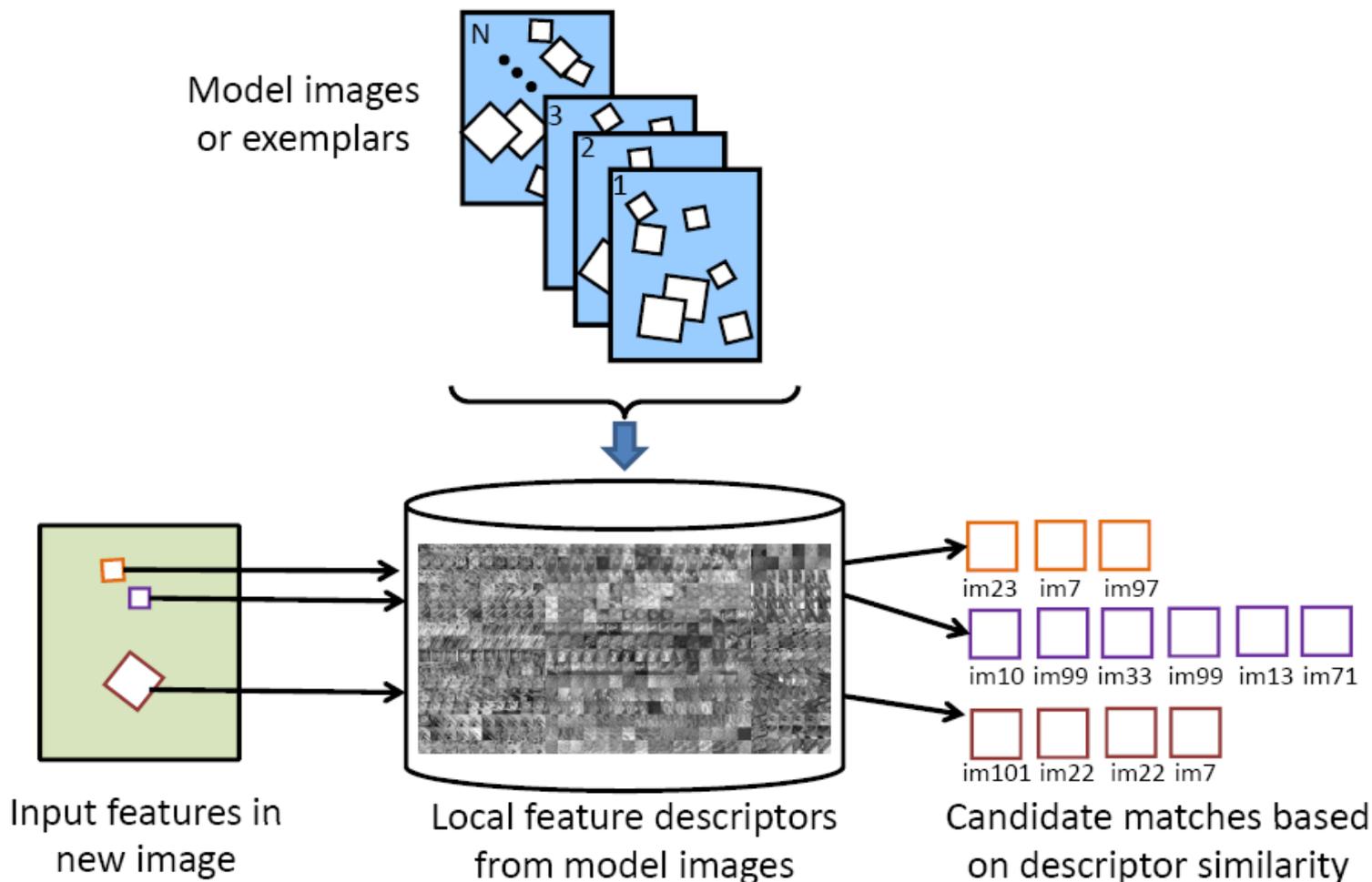
Local features for object instance recognition



D. Lowe (1999, 2004)

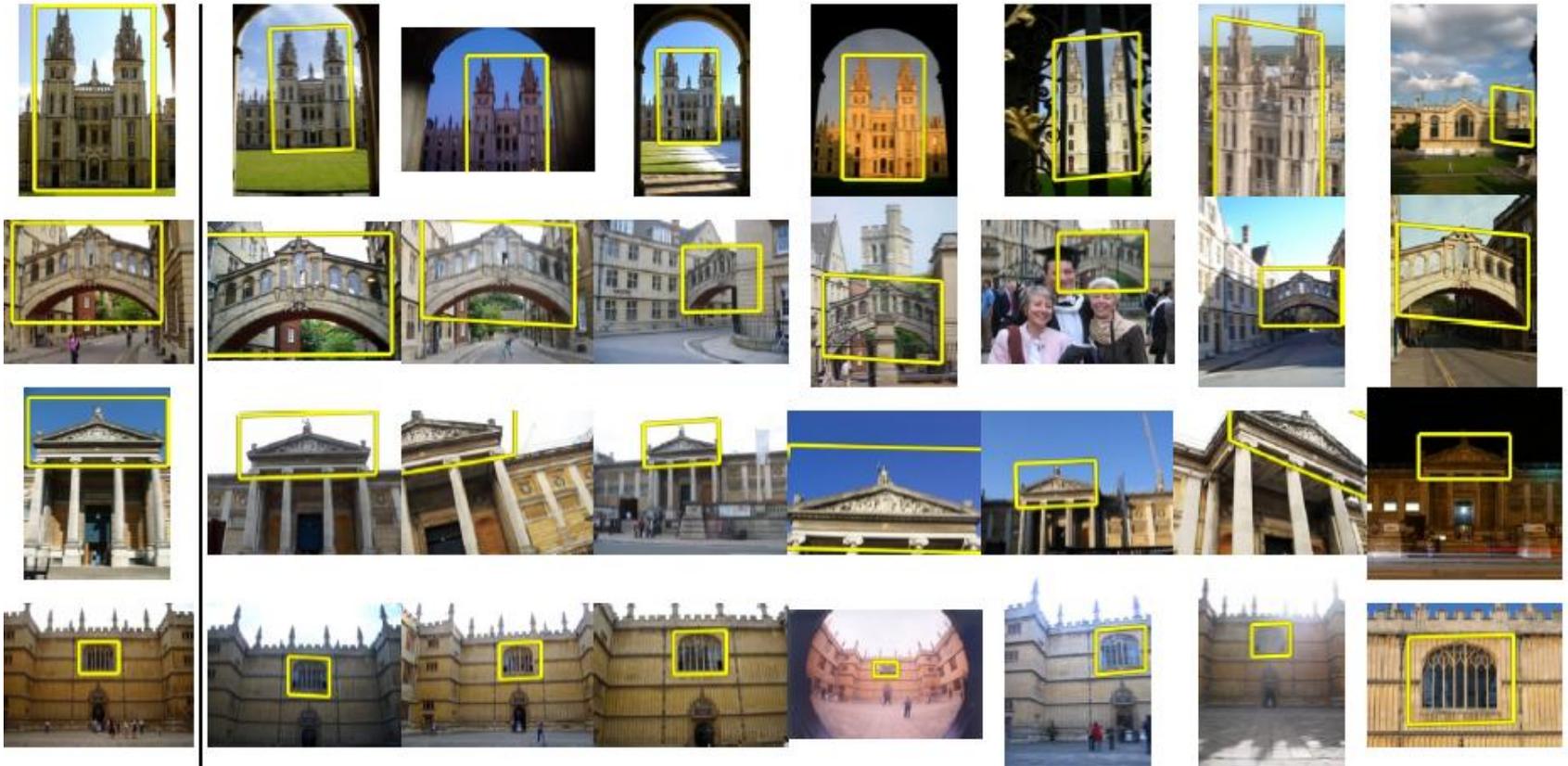
Large-scale image search

Combining local features, indexing, and spatial constraints



Large-scale image search

Combining local features, indexing, and spatial constraints

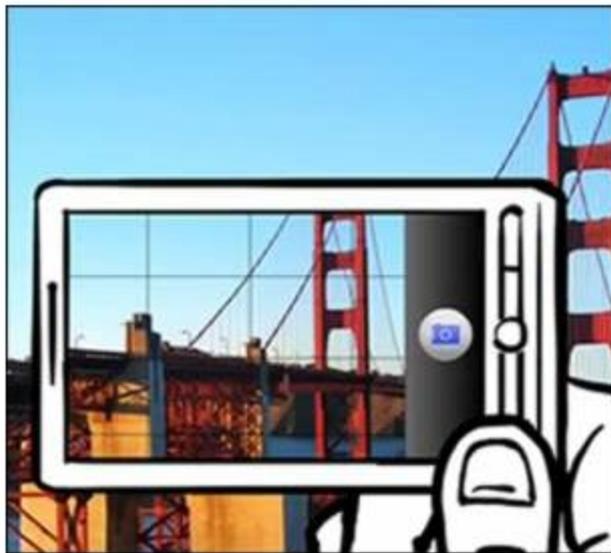


Large-scale image search

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.



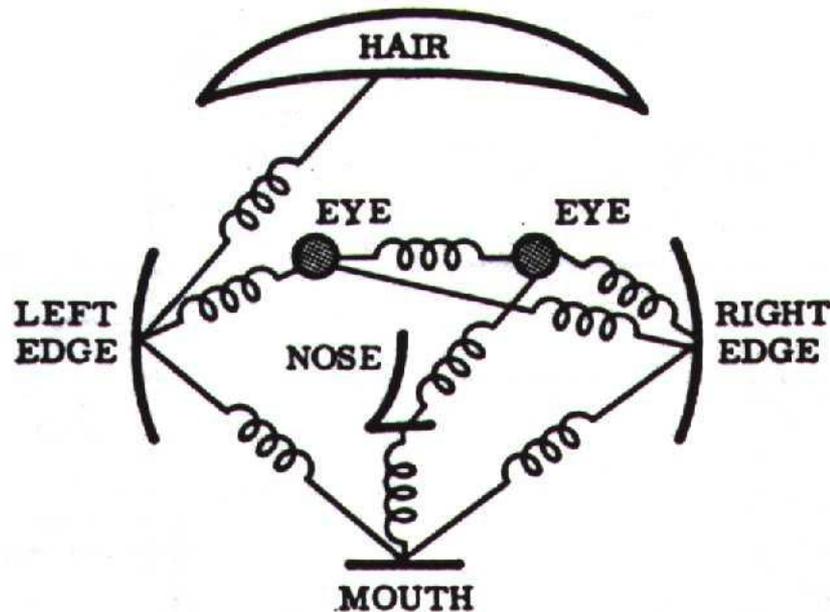
Available on phones that run Android 1.6+ (i.e. Donut or Eclair)

History of ideas in recognition

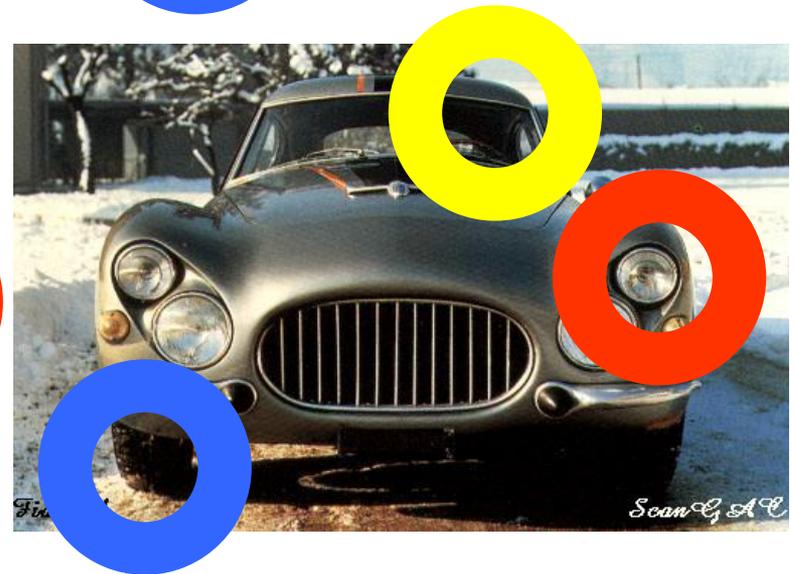
- 1960s – early 1990s: the geometric era
- 1990s: appearance-based models
- Mid-1990s: sliding window approaches
- Late 1990s: local features
- Early 2000s: parts-and-shape models

Parts-and-shape models

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



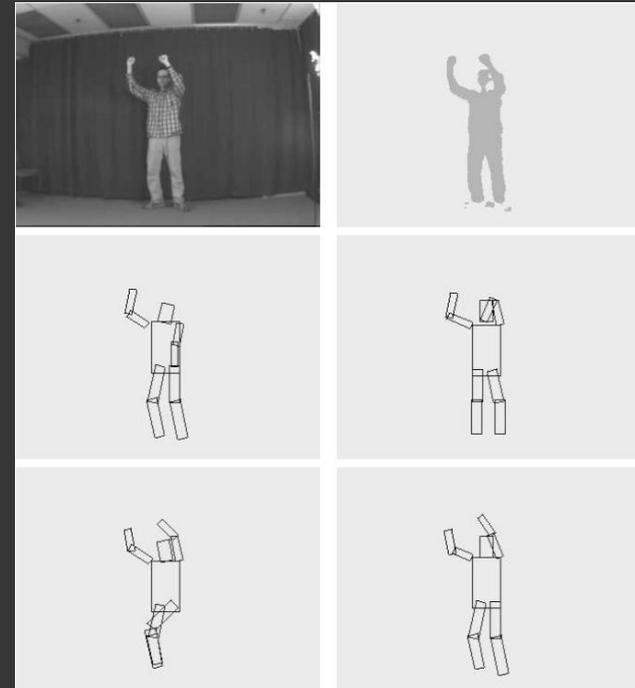
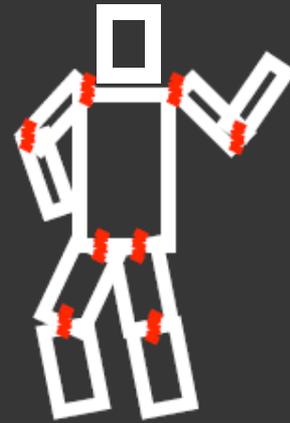
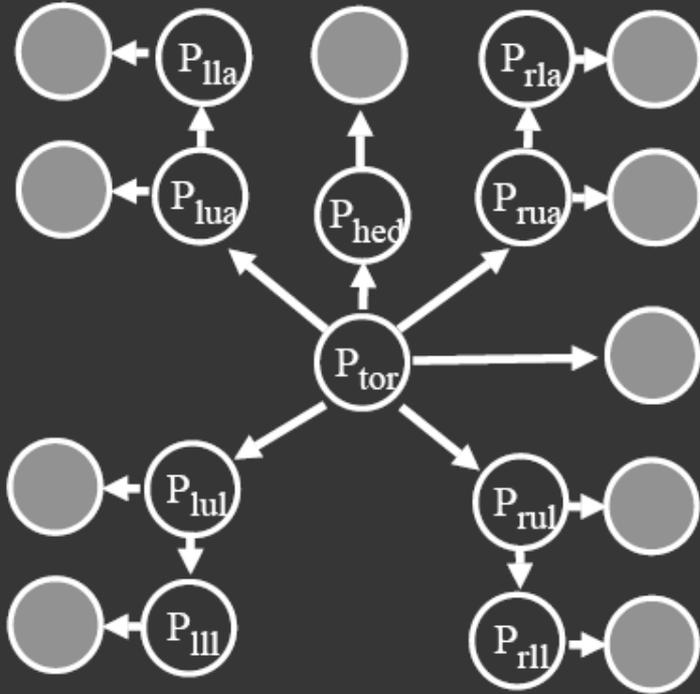
Constellation models



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

Pictorial structure model

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



$$\Pr(P_{\text{tor}}, P_{\text{arm}}, \dots | \text{Im}) \propto \prod_{i,j} \Pr(P_i | P_j) \prod_i \Pr(\text{Im}(P_i))$$

↑
↑

part geometry
part appearance

History of ideas in recognition

- 1960s – early 1990s: the geometric era
- 1990s: appearance-based models
- Mid-1990s: sliding window approaches
- Late 1990s: local features
- Early 2000s: parts-and-shape models
- Mid-2000s: bags of features

Bag-of-features models

Object

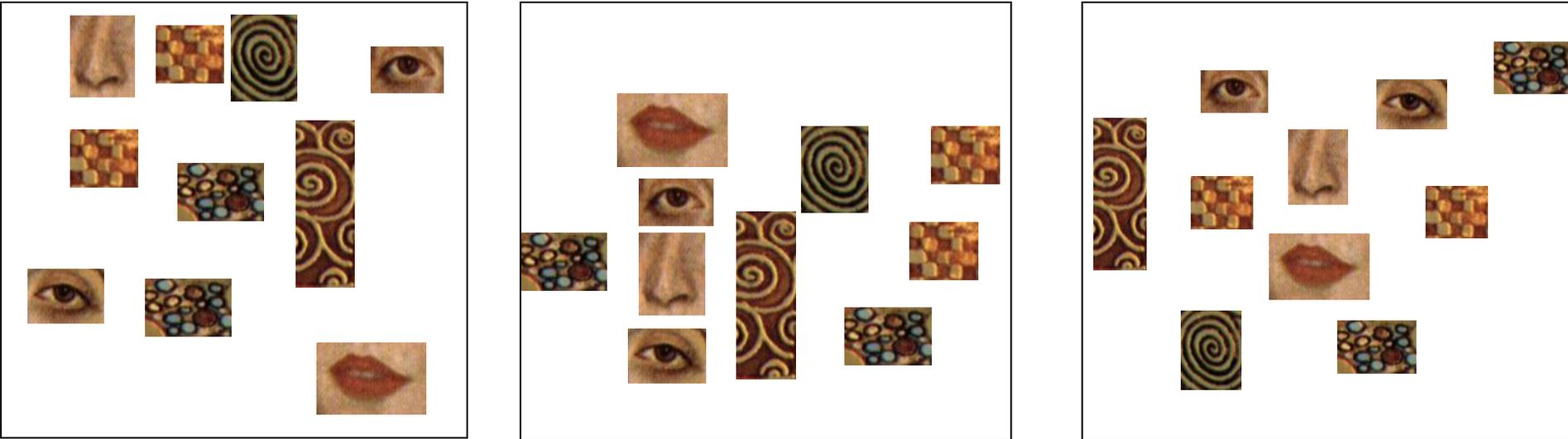


**Bag of
'words'**



Objects as texture

- All of these are treated as being the same



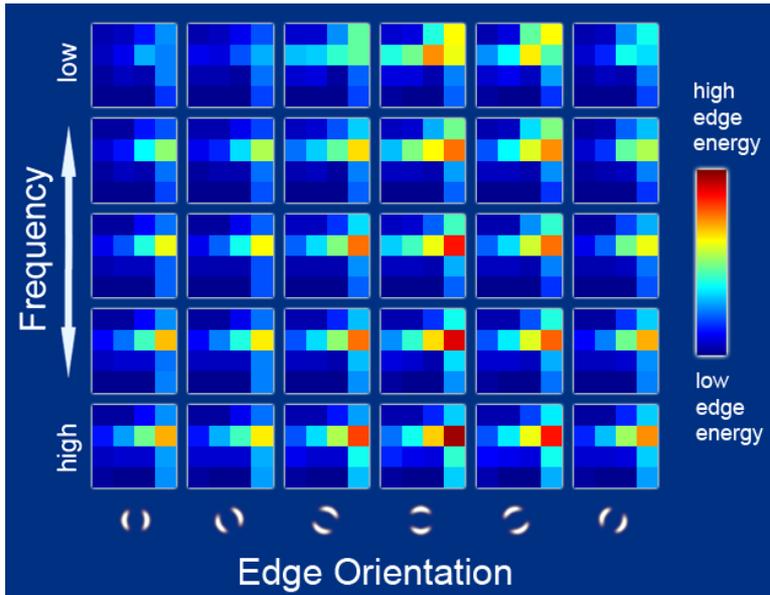
- No distinction between foreground and background: scene recognition?

History of ideas in recognition

- 1960s – early 1990s: the geometric era
- 1990s: appearance-based models
- Mid-1990s: sliding window approaches
- Late 1990s: local features
- Early 2000s: parts-and-shape models
- Mid-2000s: bags of features
- Present trends: combination of local and global methods, data-driven methods, context

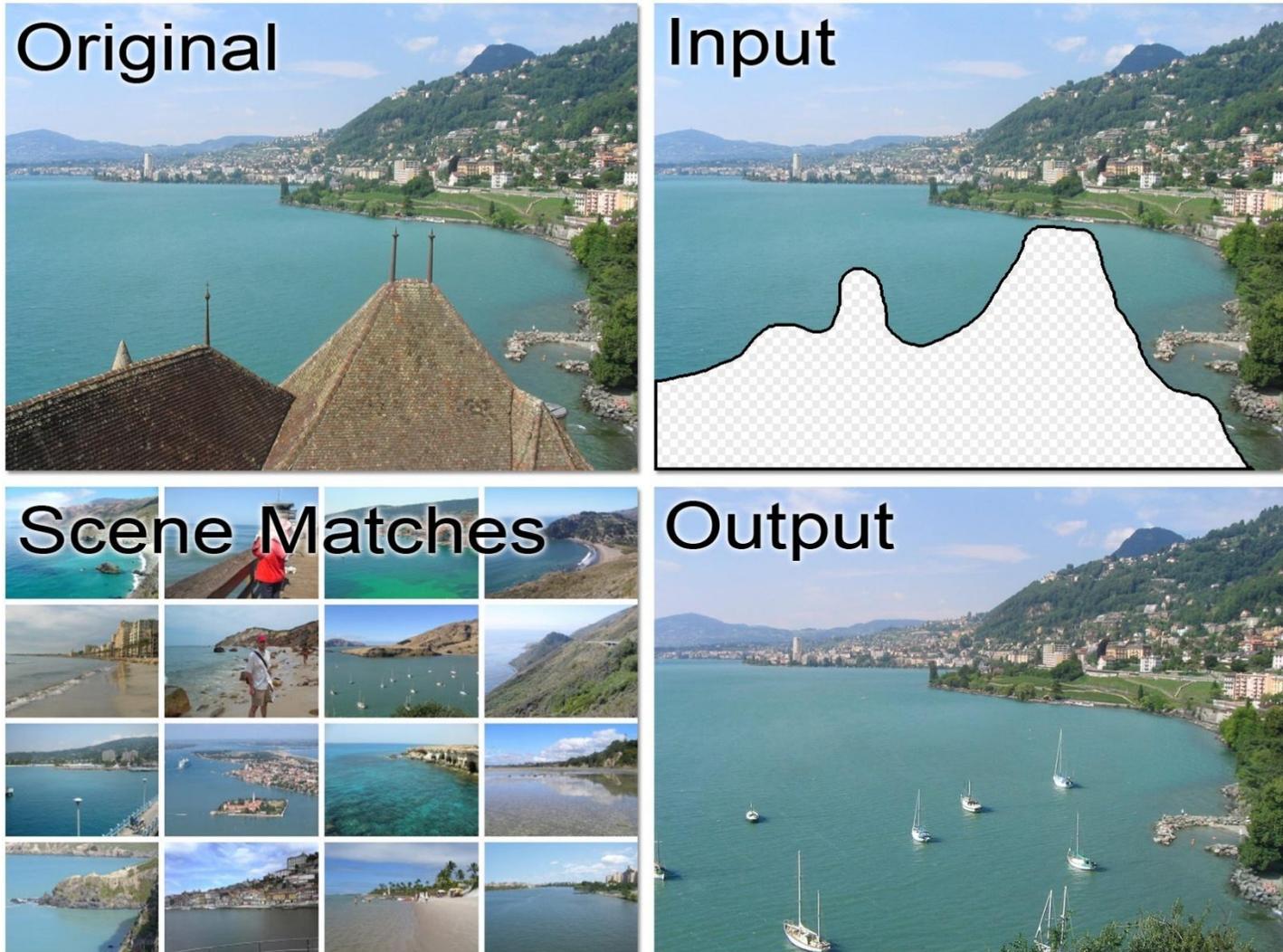
Global scene descriptors

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

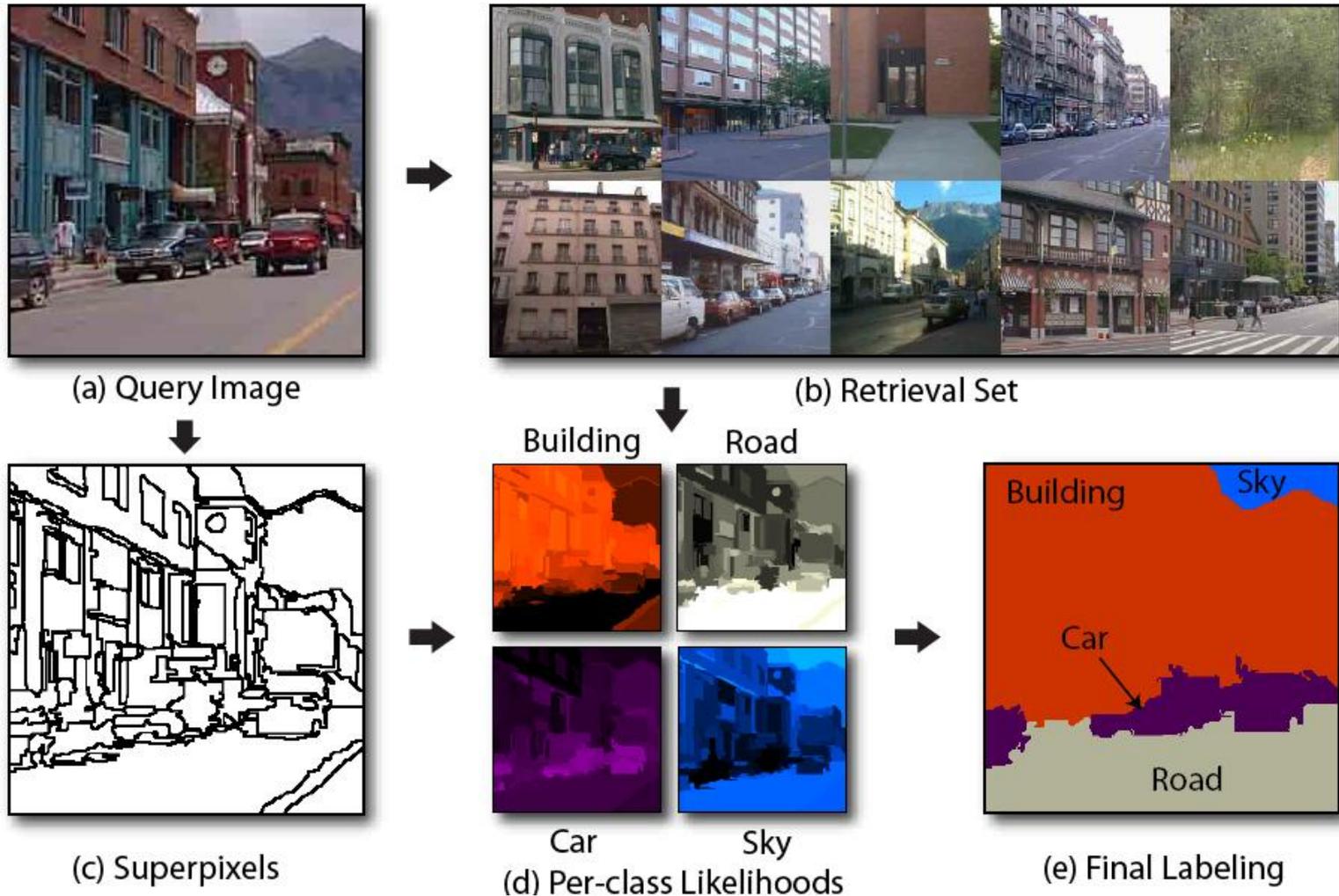


<http://people.csail.mit.edu/torralba/code/spatialenvelope/>

Data-driven methods



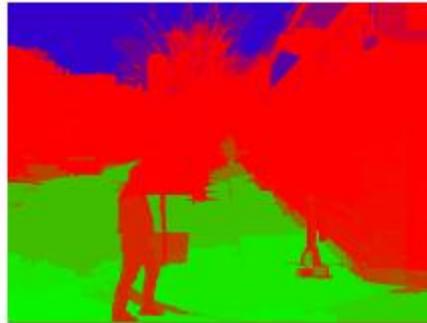
Data-driven methods



Geometric context



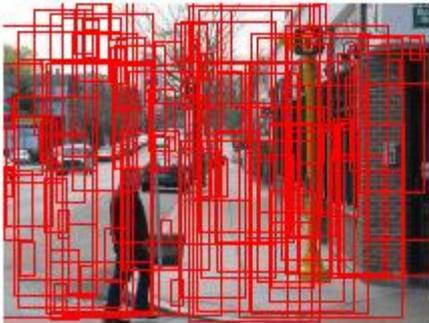
(a) Input image



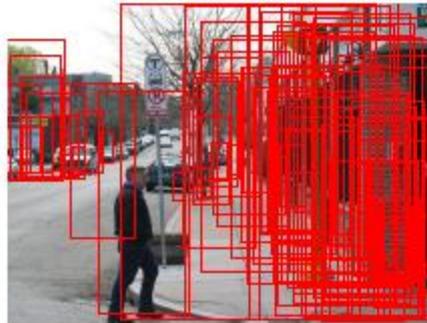
(c) Surface estimate



(e) $P(\text{viewpoint} \mid \text{objects})$



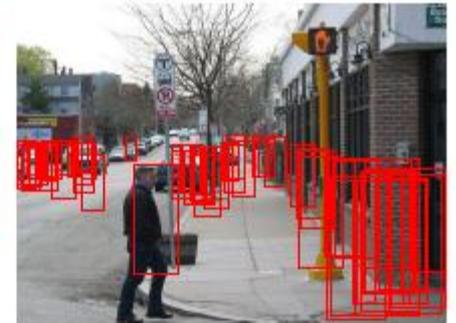
(b) $P(\text{person}) = \text{uniform}$



(d) $P(\text{person} \mid \text{geometry})$



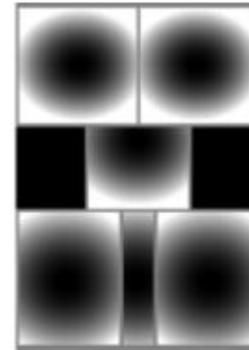
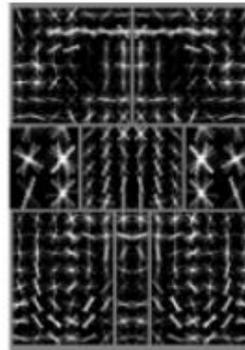
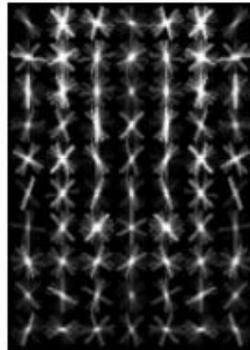
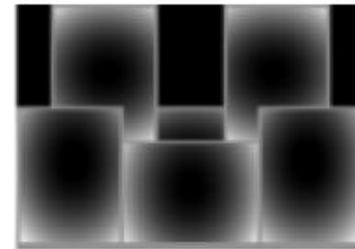
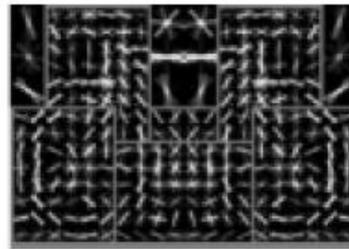
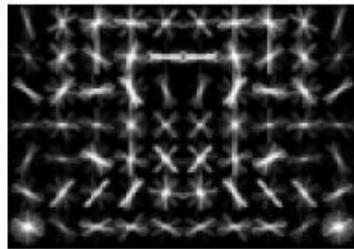
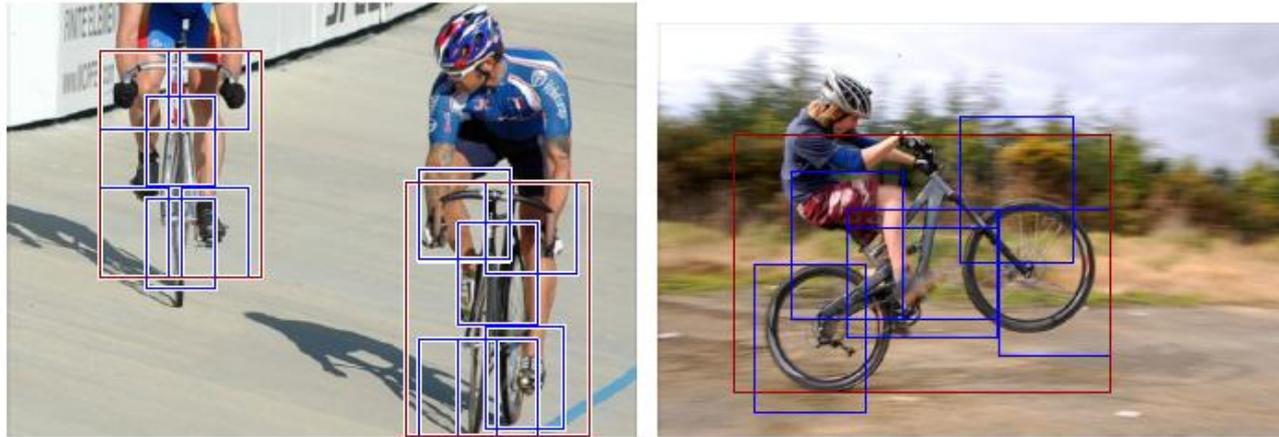
(f) $P(\text{person} \mid \text{viewpoint})$



(g) $P(\text{person} \mid \text{viewpoint, geometry})$

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

Discriminatively trained part-based models



P. Felzenszwalb, R. Girshick, D. McAllester, D. Ramanan, ["Object Detection with Discriminatively Trained Part-Based Models,"](#) PAMI 2009

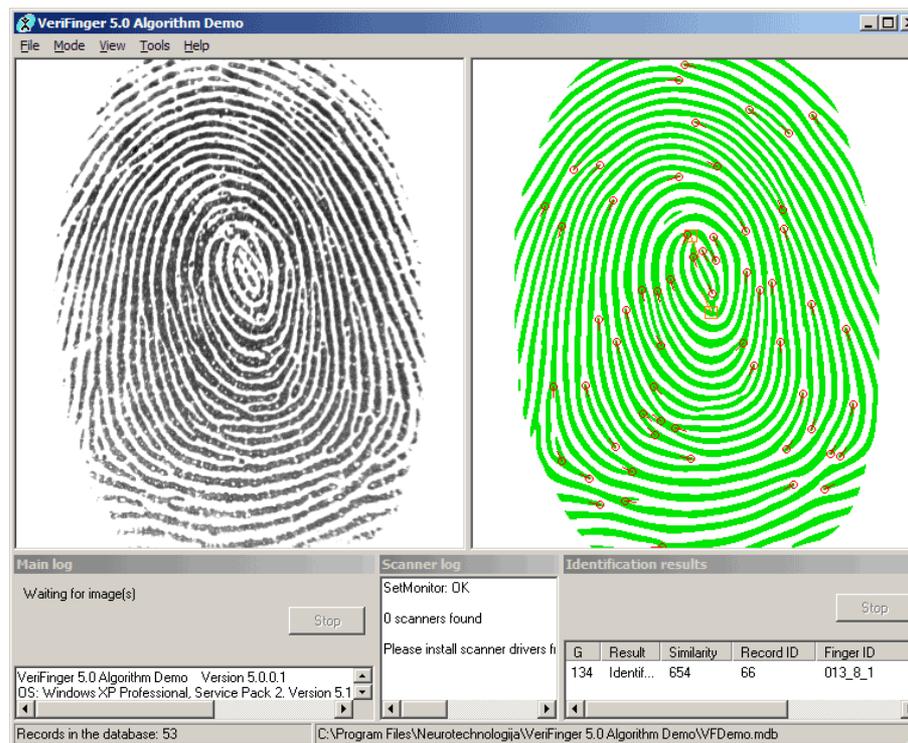
What “works” today

- Reading license plates, zip codes, checks

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

What “works” today

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



What “works” today

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



[Face priority AE] When a bright part of the face is too bright

What “works” today

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

