Object Recognition: History and Overview



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



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



Recall: Origins of computer vision

-23-4445(a-d)





(a) Original picture.

(b) Differentiated picture.

L. G. Roberts, <u>Machine Perception</u> of <u>Three Dimensional Solids</u>, Ph.D. thesis, MIT Department of Electrical Engineering, 1963.



(c) Line drawing.



Alignment: Huttenlocher & Ullman (1987)





Variability

Camera position Illumination Internal parameters



Variability

Camera position Illumination Internal parameters





Invariance to:Camera positionIlluminationInternal parameters

Duda & Hart (1972); Weiss (1987); Mundy et al. (1992-94); Rothwell et al. (1992); Burns et al. (1993) Example: invariant to similarity transformations computed from four points



Projective invariants (Rothwell et al., 1992):



General 3D objects do not admit monocular viewpoint invariants (Burns et al., 1993)

Representing and recognizing object categories is harder...



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

Recognition by components



Geons (Biederman 1987)



Generalized cylinders Ponce et al. (1989)



General shape primitives?



Forsyth (2000)

R

Zisserman et al. (1995)



Empirical models of image variability

Appearance-based techniques

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



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Eigenfaces (Turk & Pentland, 1991)



Experimental	Correct/Unknown Recognition Percentage		
Condition	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

• Can work on relatively simple patterns



• Not robust to clutter, occlusion, lighting changes





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



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

- Scale / orientation range to search over
- Speed
- Context



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Context



(f) P(person | viewpoint) (g) P(person | viewpoint, geometry)

Local features

Combining *local* appearance, spatial constraints, invariants, and classification techniques from machine learning.



Mahamud & Hebert'03

Local features for recognition of object instances









Local features for recognition of object instances





- Lowe, et al. 1999, 2003
- Mahamud and Hebert, 2000
- Ferrari, Tuytelaars, and Van Gool, 2004
- Rothganger, Lazebnik, and Ponce, 2004
- Moreels and Perona, 2005

Representing categories: Parts and Structure



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

Parts-and-shape representation

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



Bag-of-features models







Objects as texture

• All of these are treated as being the same



No distinction between foreground and background: scene recognition?

Today: A comeback for global models?

• The "gist" of a scene: Oliva & Torralba (2001)



J. Hays and A. Efros, <u>Scene Completion using</u> <u>Millions of Photographs</u>, SIGGRAPH 2007



Object Recognition by Scene Alignment

Bryan C. Russell, Antonio Torralba, Ce Liu, Rob Fergus, William T. Freeman



Input image



Goal: Recognize objects embedded in a scene









Output image with object labels transferred

Timeline of recognition

- 1965-late 1980s: alignment, geometric primitives
- Early 1990s: invariants, appearance-based methods
- Mid-late 1990s: sliding window approaches
- Late 1990s: feature-based methods
- Early 2000s: parts-and-shape models
- 2003 present: bags of features
- Present trends: combination of local and global methods, modeling context, integrating recognition and segmentation

• Reading license plates, zip codes, checks



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



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





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

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

