

CS 188: Artificial Intelligence

Lecture 24: Computer Vision

Pieter Abbeel – UC Berkeley
 Slides adapted from Trevor Darrell (and his sources)

Rough evolution of focus in recognition research

Visual Object Recognition Tutorial

1980s 1990s to early 2000s 2000-2010...

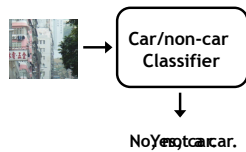
Inputs/outputs/assumptions

- What is the **goal**?
 - Say yes/no as to whether an object present in image
 - And/or:
 - Determine pose of an object, e.g. for robot to grasp
 - Categorize all objects
 - Forced choice from pool of categories
 - Bounding box on object
 - Full segmentation
 - Build a model of an object category

Scanning windows...

Detection via classification: Main idea

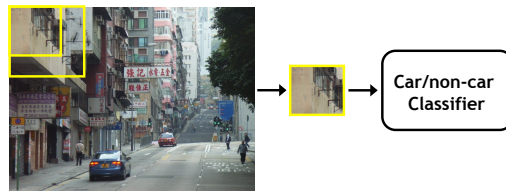
Basic component: a binary classifier



K. Grauman, B. Leibe

Detection via classification: Main idea

If object may be in a cluttered scene, slide a window around looking for it.



K. Grauman, B. Leibe

Detection via classification: Main idea

Fleshing out this pipeline a bit more, we need to:

1. Obtain training data
2. Define features
3. Define classifier

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Detection via classification: Main idea

- Consider all subwindows in an image
 - Sample at multiple scales and positions (and orientations)
- Make a decision per window:
 - “Does this contain object category X or not?”

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Feature extraction: global appearance

Simple holistic descriptions of image content

- grayscale / color histogram
- vector of pixel intensities

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Eigenfaces: global appearance description

An early appearance-based approach to face recognition

Generate low-dimensional representation of appearance with a linear subspace.

Project new images to “face space”.

Recognition via nearest neighbors in face space

Turk & Pentland, 1991

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Feature extraction: global appearance

- Pixel-based representations sensitive to small shifts

- Color or grayscale-based appearance description can be sensitive to illumination and intra-class appearance variation

Cartoon example: an albino koala

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Gradient-based representations

- Consider edges, contours, and (oriented) intensity gradients

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HOG
(one of the most widely used features)

Gradient-based representations: Histograms of oriented gradients (HoG)

Visual Object Recognition Tutorial

Map each grid cell in the input window to a histogram counting the gradients per orientation.

Code available: <http://pascal.inrialpes.fr/soft/olt/>

Dalal & Triggs, CVPR 2005 K. Grauman, B. Leibe



Slide credit: Dalal, Triggs, P. Barnum

$$\begin{bmatrix} -1 & 0 & 1 \end{bmatrix}$$

centered

$$\begin{bmatrix} -1 & 1 \end{bmatrix}$$

uncentered

$$\begin{bmatrix} 1 & -8 & 0 & 8 & -1 \end{bmatrix}$$

cubic-corrected

$$\begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$$

diagonal

$$\begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix}$$

Sobel

Slide credit: Dalal, Triggs, P. Barnum

- Histogram of gradient orientations
- Orientation
- Position
- Weighted by magnitude

Slide credit: Dalal, Triggs, P. Barnum

