

Computer Vision: Summary and Discussion

Computer Vision
CS 543 / ECE 549
University of Illinois

Derek Hoiem

Announcements

- Today is last day of regular class 😊
- Poster session next Tuesday
 - Reports due by Wed at noon
- Derek and Ian out of town Wed afternoon through Friday
 - I plan to be in office and available after class today (3:30-4:30) and Wed (10am-11am, 2pm-3pm)

Today's class

- Review of important concepts
- Some important open problems
- Feedback and course evaluation

Fundamentals of Computer Vision

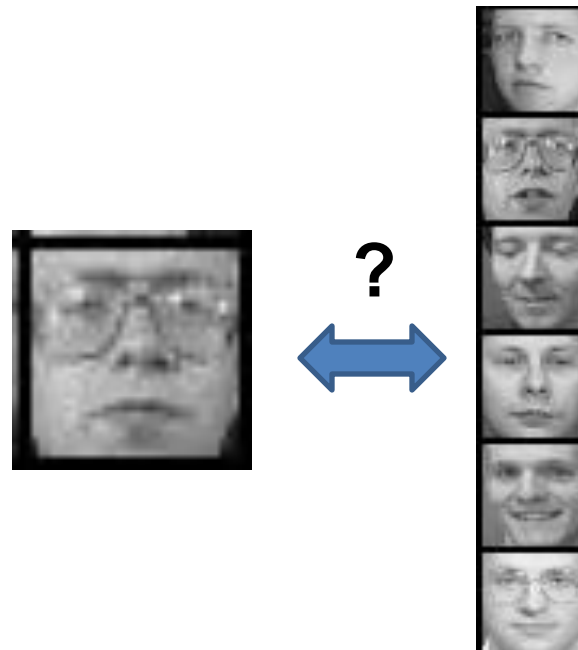
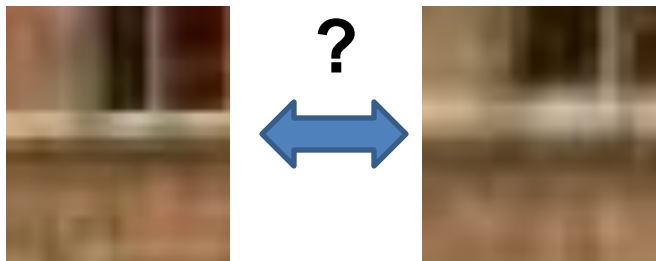
- Geometry
 - How to relate world coordinates and image coordinates
- Matching
 - How to measure the similarity of two regions
- Alignment
 - How to align points/patches
 - How to recover transformation parameters based on matched points
- Grouping
 - What points/regions/lines belong together?
- Categorization
 - What similarities are important?

Geometry

- $\mathbf{x} = \mathbf{K} [\mathbf{R} \ \mathbf{t}] \mathbf{X}$
 - Maps 3d point \mathbf{X} to 2d point \mathbf{x}
 - Rotation \mathbf{R} and translation \mathbf{t} map into 3D camera coordinates
 - Intrinsic matrix \mathbf{K} projects from 3D to 2D
- Parallel lines in 3D converge at the **vanishing point** in the image
 - A 3D plane has a vanishing line in the image
- $\mathbf{x}'^T \mathbf{F} \mathbf{x} = 0$
 - Points in two views that correspond to the same 3D point are related by the fundamental matrix \mathbf{F}

Matching

- Does this patch match that patch?
 - In two simultaneous views? (stereo)
 - In two successive frames? (tracking, flow, SFM)
 - In two pictures of the same object? (recognition)



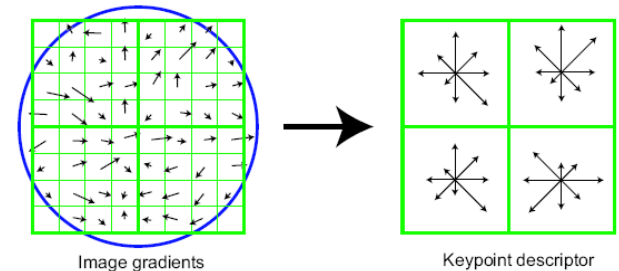
Matching

Representation: be invariant/robust to expected deformations but nothing else

- Often assume that shape is constant
 - Key cue: local differences in shading (e.g., gradients)
- Change in viewpoint
 - Rotation invariance: rotate and/or affine warp patch according to dominant orientations
- Change in lighting or camera gain
 - Average intensity invariance: oriented gradient-based matching
 - Contrast invariance: normalize gradients by magnitude
- Small translations
 - Translation robustness: histograms over small regions

But can one representation do all of this?

- SIFT: local normalized histograms of oriented gradients provides robustness to in-plane orientation, lighting, contrast, translation
- HOG: like SIFT but does not rotate to dominant orientation



Alignment of points

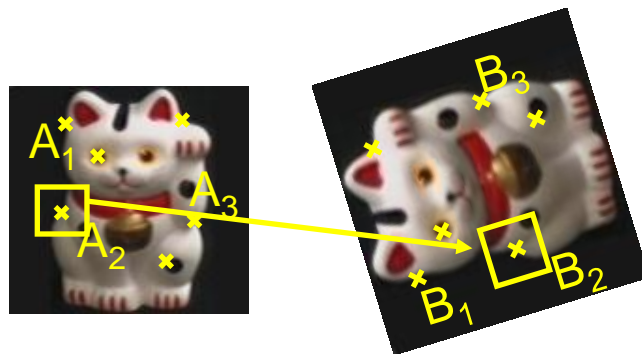
Search: efficiently align matching patches

- Interest points: find repeatable, distinctive points
 - Long-range matching: e.g., wide baseline stereo, panoramas, object instance recognition
 - Harris: points with strong gradients in orthogonal directions (e.g., corners) are precisely repeatable in x-y
 - Difference of Gaussian: points with peak response in Laplacian image pyramid are somewhat repeatable in x-y-scale
- Local search
 - Short range matching: e.g., tracking, optical flow
 - Gradient descent on patch SSD, often with image pyramid
- Windowed search
 - Long-range matching: e.g., recognition, stereo w/ scanline

Alignment of sets

Find transformation to align matching sets of points

- Geometric transformation (e.g., affine)
 - Least squares fit (SVD), if all matches can be trusted
 - Hough transform: each potential match votes for a range of parameters
 - Works well if there are very few parameters (3-4)
 - RANSAC: repeatedly sample potential matches, compute parameters, and check for inliers
 - Works well if fraction of inliers is high and few parameters (4-8)
- Other cases
 - Thin plate spline for more general distortions
 - One-to-one correspondence (Hungarian algorithm)



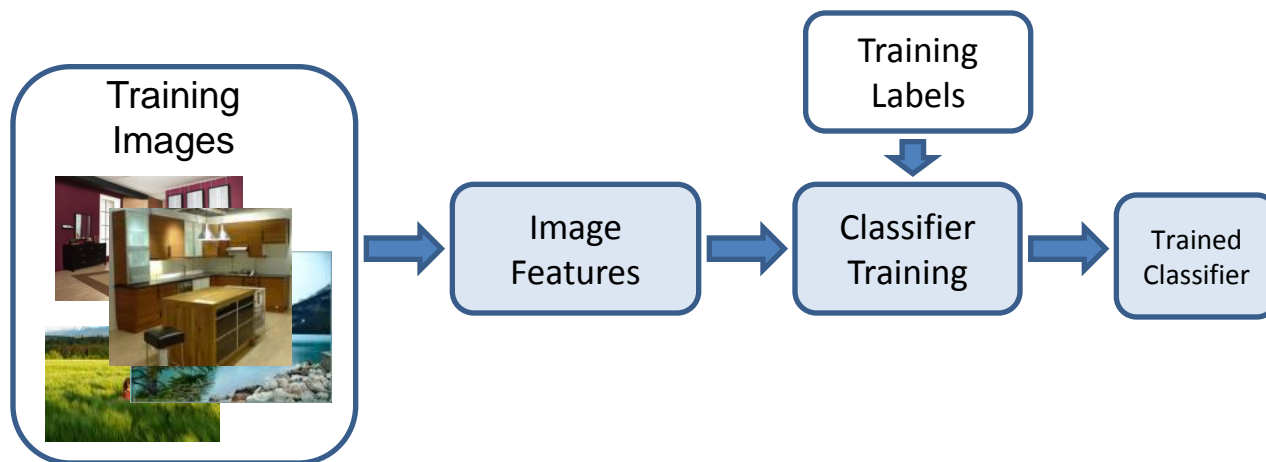
Grouping

- Clustering: group items (patches, pixels, lines, etc.) that have similar appearance
 - Discretize continuous values; typically, represent points within cluster by center
 - Improve efficiency: e.g., cluster interest points before recognition
 - Summarize data
- Segmentation: group pixels into regions of coherent color, texture, motion, and/or label
 - Mean-shift clustering
 - Watershed
 - Graph-based segmentation: e.g., MRF and graph cuts
- EM, mixture models: probabilistically group items that are likely to be drawn from the same distribution, while estimating the distributions' parameters

Categorization

Match objects, parts, or scenes that may vary in appearance

- Categories are typically defined by human and may be related by function, cost, or other non-visual attributes
- Key problem: what are important similarities?
 - Can be learned from training examples



Categorization

Representation: ideally should be compact, comprehensive, direct

- Histograms of quantized interest points (SIFT, HOG), color, texture
 - Typical for image or region categorization
 - Degree of spatial encoding is controllable by using spatial pyramids
- HOG features at specified position
 - Often used for finding parts or objects

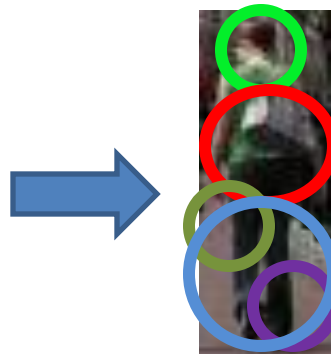
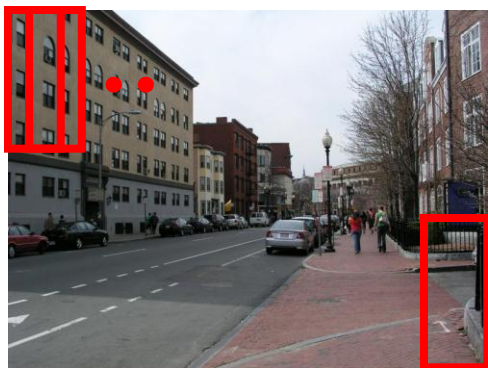
Object Categorization

Search by Sliding Window Detector

- May work well for rigid objects



- Key idea: simple alignment for simple deformations



Object or
Background?

Object Categorization

Search by Parts-based model

- Key idea: more flexible alignment for articulated objects
- Defined by models of **part appearance**, **geometry** or spatial layout, and **search algorithm**



Vision as part of an intelligent system



3D Scene

Feature
Extraction

Texture

Color

Optical
Flow

Stereo
Disparity

Grouping

Surfaces

Bits of
objects

Sense of
depth

Motion
patterns

Interpretation

Objects

Agents
and goals

Shapes and
properties

Open
paths

Words

Action

Walk, touch, contemplate, smile, evade, read on, pick up, ...

Important open problems

Computer vision is potentially worth major \$\$\$, but there are major challenges to overcome first.

- Driver assistance
 - MobileEye received >\$100M in funding from Goldman Sachs
- Entertainment (Kinect, movies, etc.)
 - Intel is spending \$100M for visual computing over next five years
- Security
 - Potential for billions of deployed cameras
- Robot workers
- Many more

Important open problems

Object category recognition: where is the cat?



Important open problems

Object category recognition: where is the cat?



Important questions:

- How can we better align two object instances?
- How do we identify the important similarities of objects within a category?
- How do we tell if two patches depict similar shapes?

Important open problems

Object representation: what is it?



Important open problems

Object representation: what is it?



Important questions:

- How can we pose recognition so that it lets us deal with new objects?
- What do we want to predict or infer, and to what extent does that rely on categorization?
- How do we transfer knowledge of one type of object to another?

Important open problems

- Spatial understanding: what is it doing? Or how do I do it?



Important open problems

- Spatial understanding: what is it doing? Or how do I do it?



Important questions:

- What are good representations of space for navigation and interaction? What kind of details are important?
- How can we combine single-image cues with multi-view cues?

Important open problems

- Algorithms: works pretty well → perfect
 - E.g., stereo: top of wish list from Pixar guy Micheal Kass

Good directions:

- Incorporate higher level knowledge

Important open problems

- How should we adjust vision systems to solve particular tasks?

Important open problems

- Can we build a “core” vision system that can easily be extended to perform new tasks or even learn on its own?
 - What kind of representations might allow this?
 - What should be built in and what should be learned?

If you want to learn more...

- Read lots of papers: IJCV, PAMI, CVPR, ICCV, ECCV, NIPS
- Helpful topics for classes
 - David Forsyth's optimization
 - Classes in machine learning or pattern recognition
 - Statistics, graphical models
 - Seminar-style paper-reading classes
- Just implement stuff, try demos, see what works

Feedback: very important

- My custom form
- ICES forms

See you next week!

- Project posters on Tuesday
 - Pizza provided
 - Posters: 24" wide x 32" tall