

# CMPSCI 370: Introduction to Computer Vision

University of Massachusetts, Amherst  
January 19, 2016

Instructor: Subhransu Maji

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## Administrivia

- **Lectures**
  - Tuesday/Thursday, 11:30 - 12:45, Hasbrouck 113
  - Honor's section: Tuesday 4:00 - 5:00, CS 142
- **Instructor:** Subhransu Maji
- **Office hours:** Monday, 3:00 - 5:00, CS 274
- **Website:** <http://www-edlab.cs.umass.edu/~smaji/cmpsci370/>
  - News, lecture slides, etc (check regularly)
  - Homework submission via Moodle

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## Administrivia

- **Grading policy:**
  - 370: homework (60%), mid-term (15%), final (25%)
  - 370HH: homework (45%), mid-term (10%), final (15%), project (20%)
- **Homework**
  - 5 in total (expect one every two weeks)
  - First one will be posted this Thursday
- **Course textbooks (recommended)**
  - Richard Szeliski, Computer Vision: Algorithms and Applications (available [online](#) as pdf) - readings will be from this
- **Necessary background:** Linear algebra, calculus, probability, programming in Matlab (image toolbox needed)

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## Homework #0

1. Figure out a way to run Matlab
    - Obtain a student copy (Matlab suite 99\$)
    - Your lab machines might have it
  2. Learn how to program in Matlab
    - Plenty of online resources (the course website lists some)
- **Alternatives:** Python, Octave, JAVA, C++, ....
  - **Question:** How many of you are familiar with Matlab?

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## Before we start ...

- Are there any questions?

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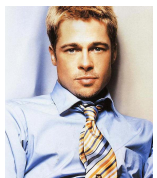
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## Why Vision?

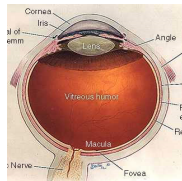
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## Why Vision? Light!



It is how we see other people, navigate our environment, communicate ideas, entertain, and **measure** the world around us.



Source: A. Berg

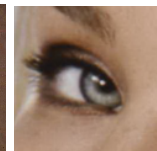
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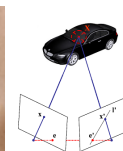
## Why is light good for measurement?



Microscopy



Surveillance



3D Analysis / Navigation



Remote Sensing

- Plentiful, sometimes free
- Interacts with many things, but not too many
- Goes generally straight over distance
- Very small → high spatial resolution
- Fast, but not too fast → time of flight sensors
- Easy to detect → cameras work, are cheap
- Comes in many flavors ( wavelengths )



Source: A. Berg

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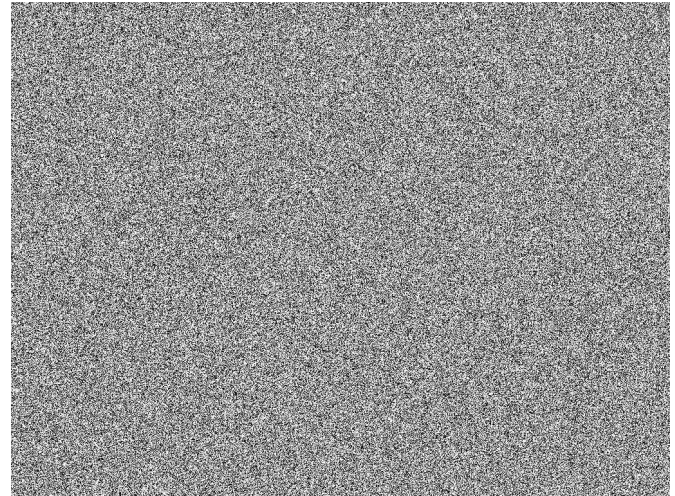
## The goal of computer vision

Extract properties of the world from visual data  
(i.e., measurements of light)

We are remarkably good at this!

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## An experiment ...#1



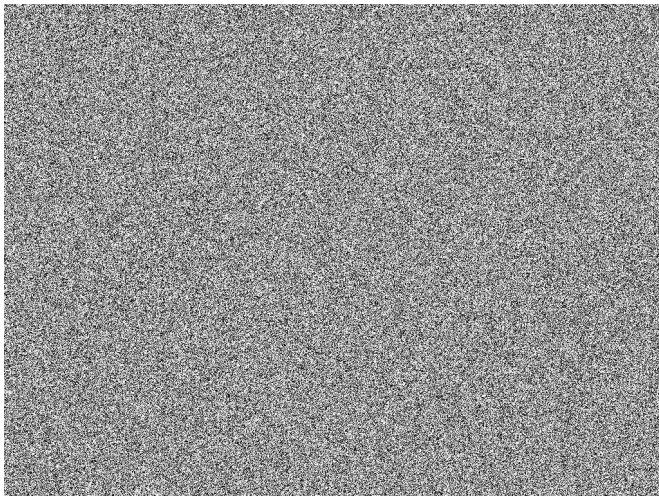
animal or not?

10

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## An experiment ...#2

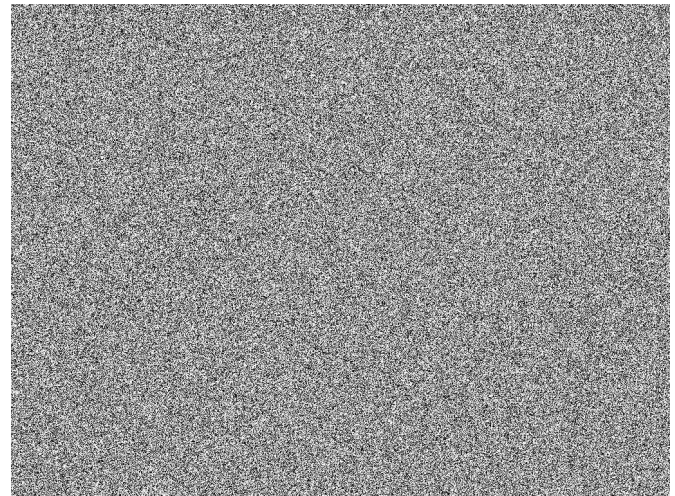


animal or not?

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## An experiment ...#3



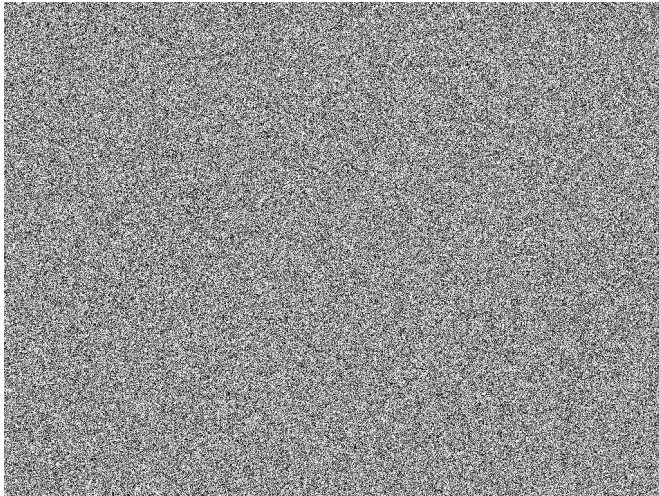
animal or not?

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## An experiment ...#4

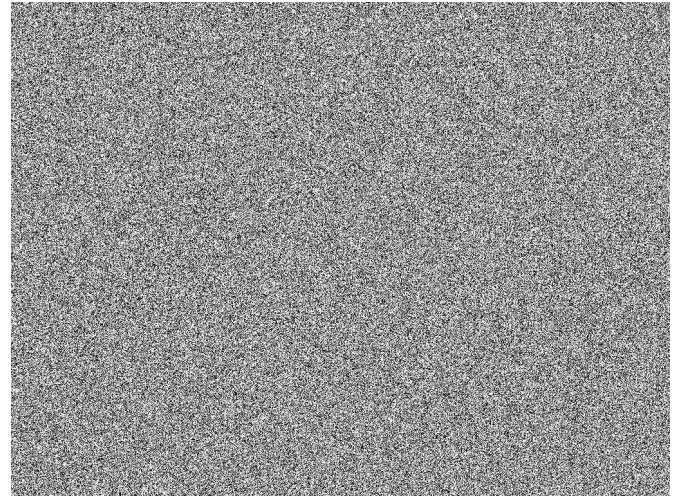


animal or not?

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## An experiment ...#5

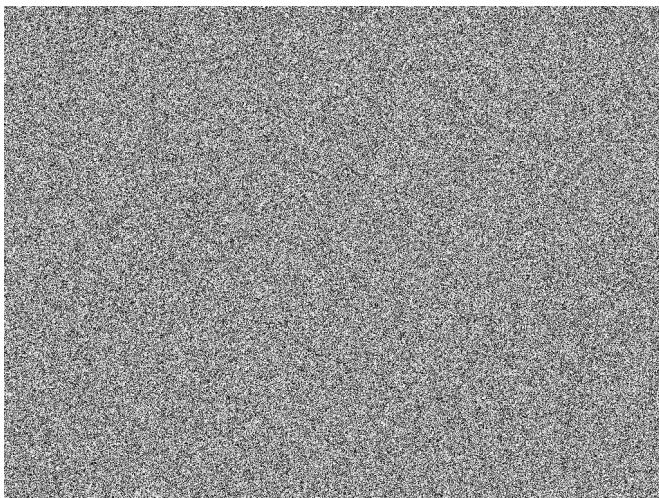


animal or not?

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## An experiment ...#6



animal or not?

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## The images ...



#1



#2



#3



#4



#5



#6

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## Human vision

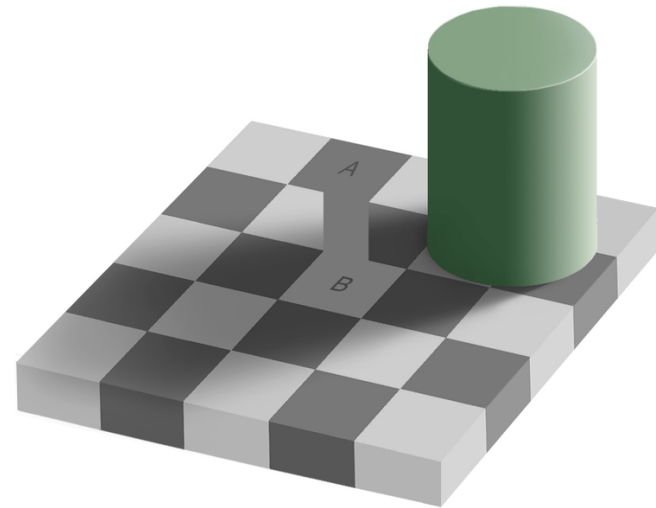
- Amazingly good, fast and accurate
- Sometimes wrong, but often not in doubt
- Huge amount of bandwidth to the brain is visual data
- Large amount of the brain seems to be for processing visual data
- Vision is difficult!

Source: A. Berg

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## But we make mistakes ...

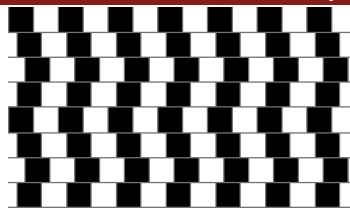


Checker shadow illusion - Edward H. Adelson

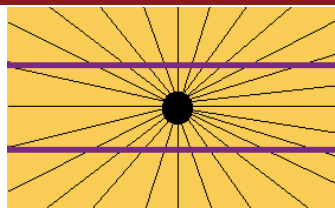
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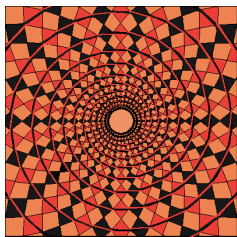
## Other optical illusions



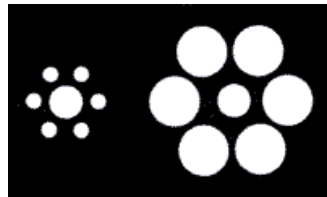
Are the horizontal lines parallel?



Are the purple lines straight?



Is this a spiral?



is the left circle (in the center) bigger?

Are these failures of our vision system?

<http://www.illusions.org> 19

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## Vision as inverse of graphics

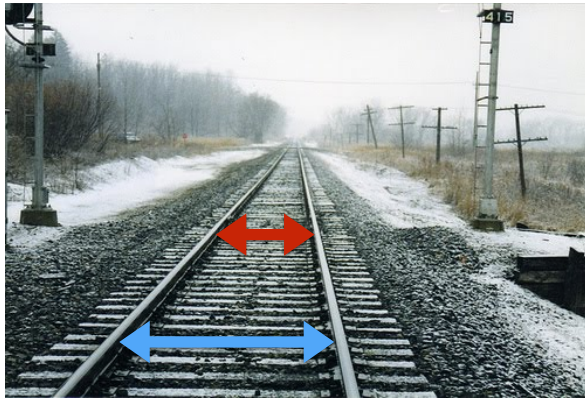
- Many possibilities — how do we solve this ambiguity?
- Images are confusing, but they also reveal the structure of the world through numerous cues
- Our job is to interpret the cues!



(following slides from J. Koenderink) 20

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## Cues: Linear perspective



Parallel lines  
merge at the  
horizon

<http://kalisdigitalphotos.blogspot.com>

Analyzing parallel lines to estimate space

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## Cues: Aerial (Atmospheric) perspective



Scattering of skylight by  
particles in the air adds  
to the luminosity

Photo by Éole Wind

As the distance of the object from the viewer *increases*, the  
contrast between the object and its background *decreases*.

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## Cues: Occlusion ordering



Chicago loop, image source: [wikipedia](https://en.wikipedia.org/wiki/Chicago_Skyline)

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## Cues: Texture gradient



Gustave Caillebotte. Paris Street, Rainy Day, 1877, Art Institute of Chicago

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## Cues: Shading and Lighting



"The four seasons" sculpture set

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## Many other cues ...

- **Motion parallax:** how things move relative to each other as we move. Objects near us move more than objects far away. Also provides *grouping* cues.
- **Familiar size:** Size of known things, e.g. faces gives us an estimate of the depth.
- **Defocus blur:** Far away objects are blurrier than nearer. Commonly used in photographs to create a perception of depth.
- **Elevation:** Distance from the horizon. Objects closer to the horizon are perceived to be farther.

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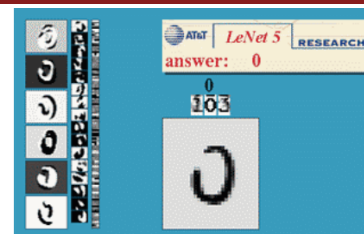
## The study of computer vision

- Lots of tasks: detection, classification, segmentation, pose estimation, depth estimation, etc.
- Problems are often ill-posed. Most of the hard work is in crisply defining the problem you wish to solve.
- It is hard, ad-hoc. There are few theorems, but we rely on those from many other areas: optics, geometry, physics, etc.
- You are in good company:
  - Euclid, Alhazen, da Vinci, Kepler, Galileo, Descartes, Newton, Huygens, Maxwell, Helmholtz, Mach, Herring, Cajal, Minkowski, Hubel & Wiesel, Wald
- If that is not enough, there are many applications

(following slides from Charless Flowkes) 27

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## Optical character recognition (OCR)



Digit recognition  
[yann.lecun.com](http://yann.lecun.com)



License plate readers  
(google street view)



Automatic cheque readers  
(Most bank ATMs)



Sudoku grabber  
<http://sudokugrab.blogspot.com/>

Source: S. Seitz, N. Snavely

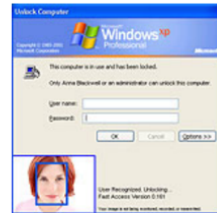
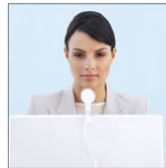
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## Biometrics



Fingerprint scanners are now on many new laptops and other devices



Face recognition systems are beginning to appear more widely  
<http://www.sensiblevision.com>

Source: S. Seitz 29

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## Face detection



Face detection is on many cameras these days

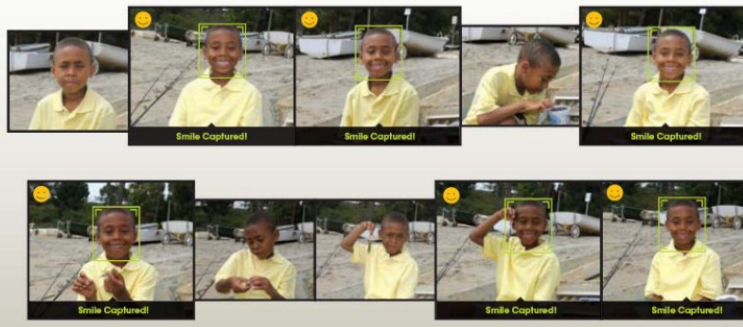
Source: S. Seitz 30

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## Smile detection

### The Smile Shutter flow

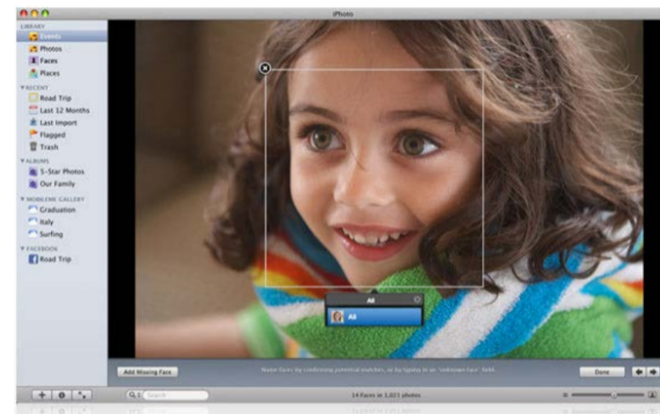
Imagine a camera smart enough to catch every smile! In Smile Shutter Mode, your Cyber-shot® camera can automatically trip the shutter at just the right instant to catch the perfect expression.



Source: S. Seitz 31

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## Face recognition



<http://www.apple.com/ilife/iphoto>

Source: S. Seitz 32

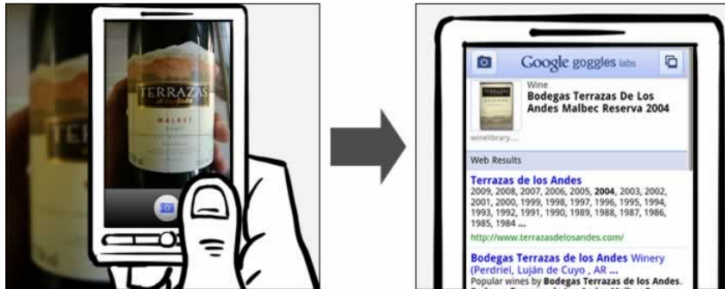
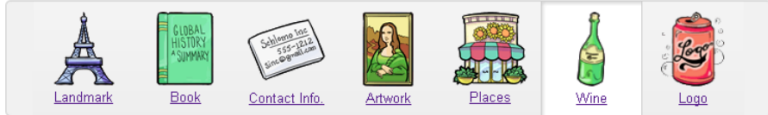
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# Instance recognition

## Google Goggles in Action

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



Source: S. Seitz

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# Automotive safety



- **Mobileye** : Vision systems on high end BMW, GM, Volvo models
  - Pedestrian collision warning
  - Forward collision warning
  - Lane departure warning
  - Headway monitoring and warning

Source: A. Shashua, S. Seitz

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# Self-driving cars



Source: L. Lazebnik

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# Interactive interfaces

## Microsoft Kinect depth sensors



Source: L. Lazebnik

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## Large-scale 3D reconstruction



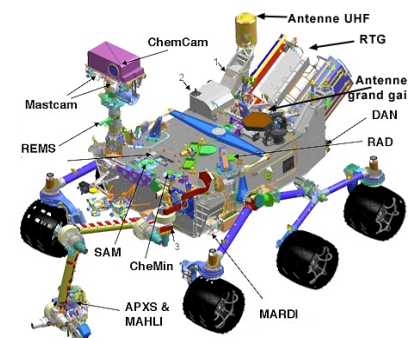
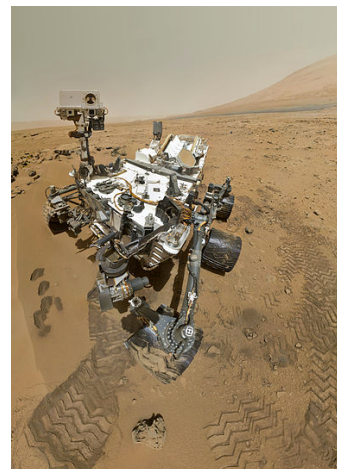
Photo Tourism: Exploring Photo Collections in 3D

[YouTube link](#)

Source: S. Seitz, N. Snavely 37

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## Vision for robotics, space exploration



NASA's Curiosity Rover has 17 cameras as a part of its sensing system

[http://en.wikipedia.org/wiki/Curiosity\\_\(rover\)](http://en.wikipedia.org/wiki/Curiosity_(rover))

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## What this course is about?

- Course overview
  - Early vision:** image formation, sensing, light and shading, filtering
  - Mid-level vision :** grouping, perceptual organization
  - Multi-view geometry**
  - Recognition**
  - Additional topics** (time permitting)
- **Goal:** To develop vision researchers. You can come up with a reasonable solution to various vision problems (and implement it yourself).
- We are not going to cover:
  - Graphics: Physics of light transport, material properties, rendering
  - Computational photography: design of sensing devices, etc
  - How the human vision system works

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## I. Early vision

- Basic image formation and processing

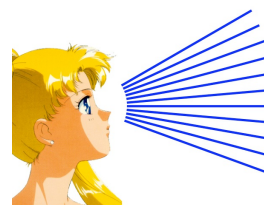


image formation

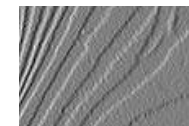
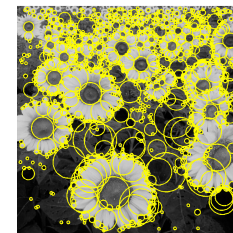


image filtering



feature extraction, key-point detection

Source: L. Lazebnik 40

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## II. Mid-level vision

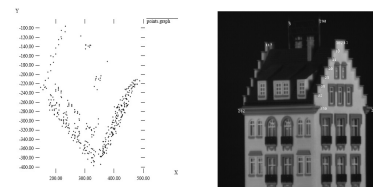
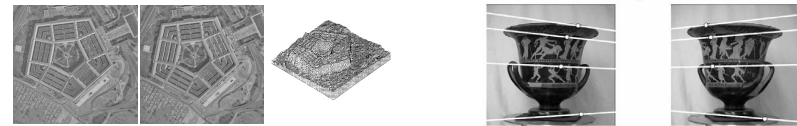
- Model fitting and grouping



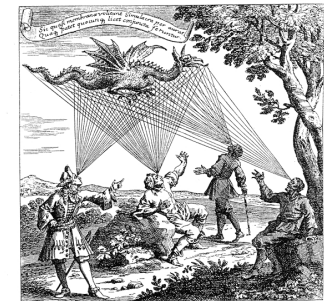
Source: L. Lazebnik 41

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## III. Multi-view geometry



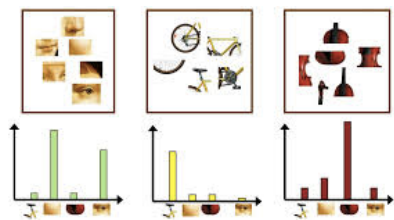
structure from motion



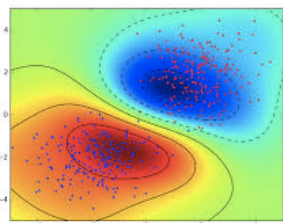
Source: L. Lazebnik 42

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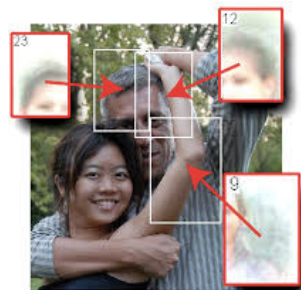
## IV. Recognition



bag-of-words models



learning

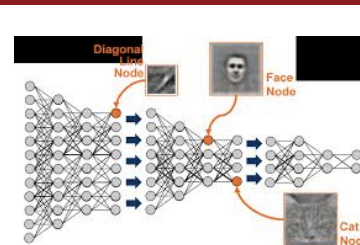


part-based models

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## V. Additional topics



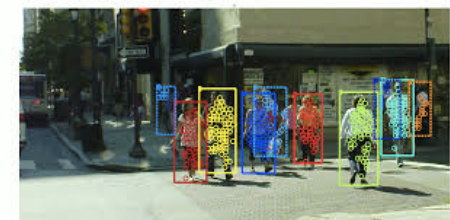
Deep learning



Optical flow



Human-centric vision



Tracking

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## For next class ...

- Familiarize yourself with MATLAB (more information is on the course page)
  - **Student copy** is 99\$ from Matlab's page
  - **UMASS IT** (100% free): <https://www.it.umass.edu/support/software>
- Readings:
  - **The speed of processing in the human visual system**, Thorpe et al., Letters to Nature, 1996
  - Chapter 1 in RS textbook