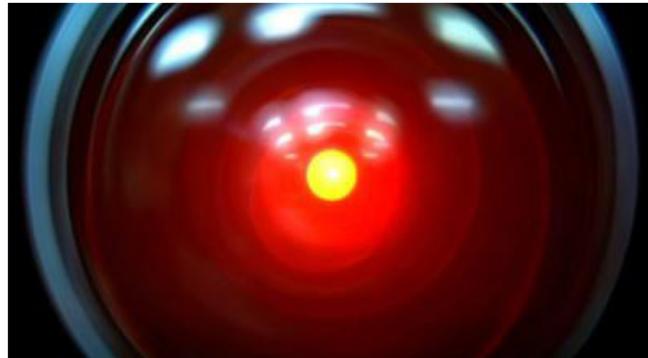
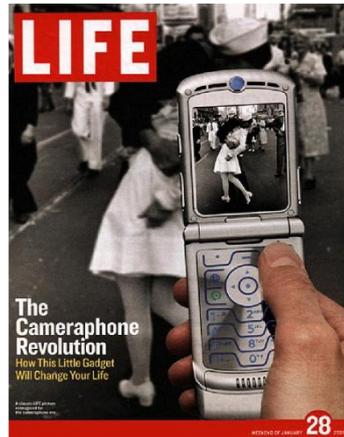
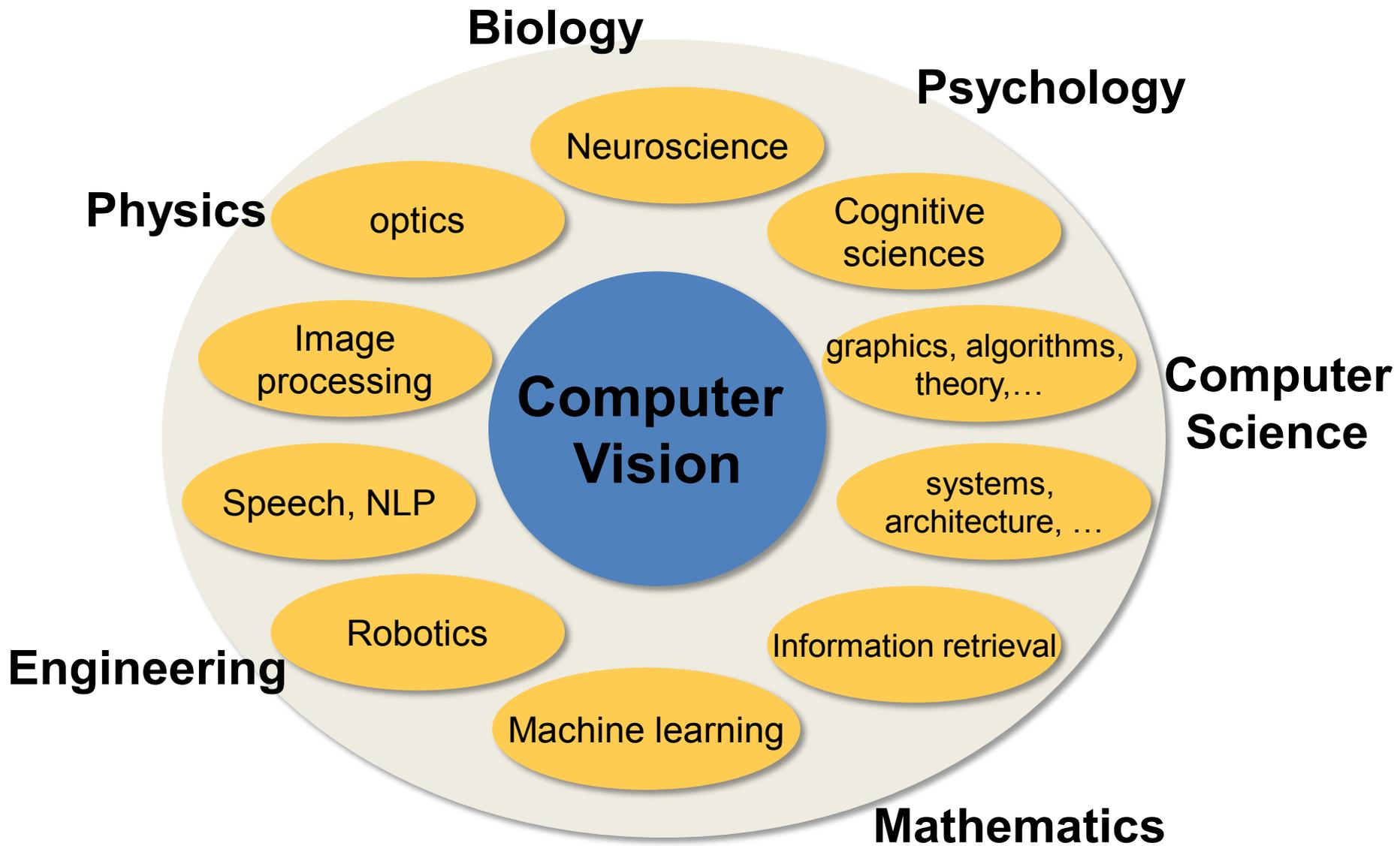


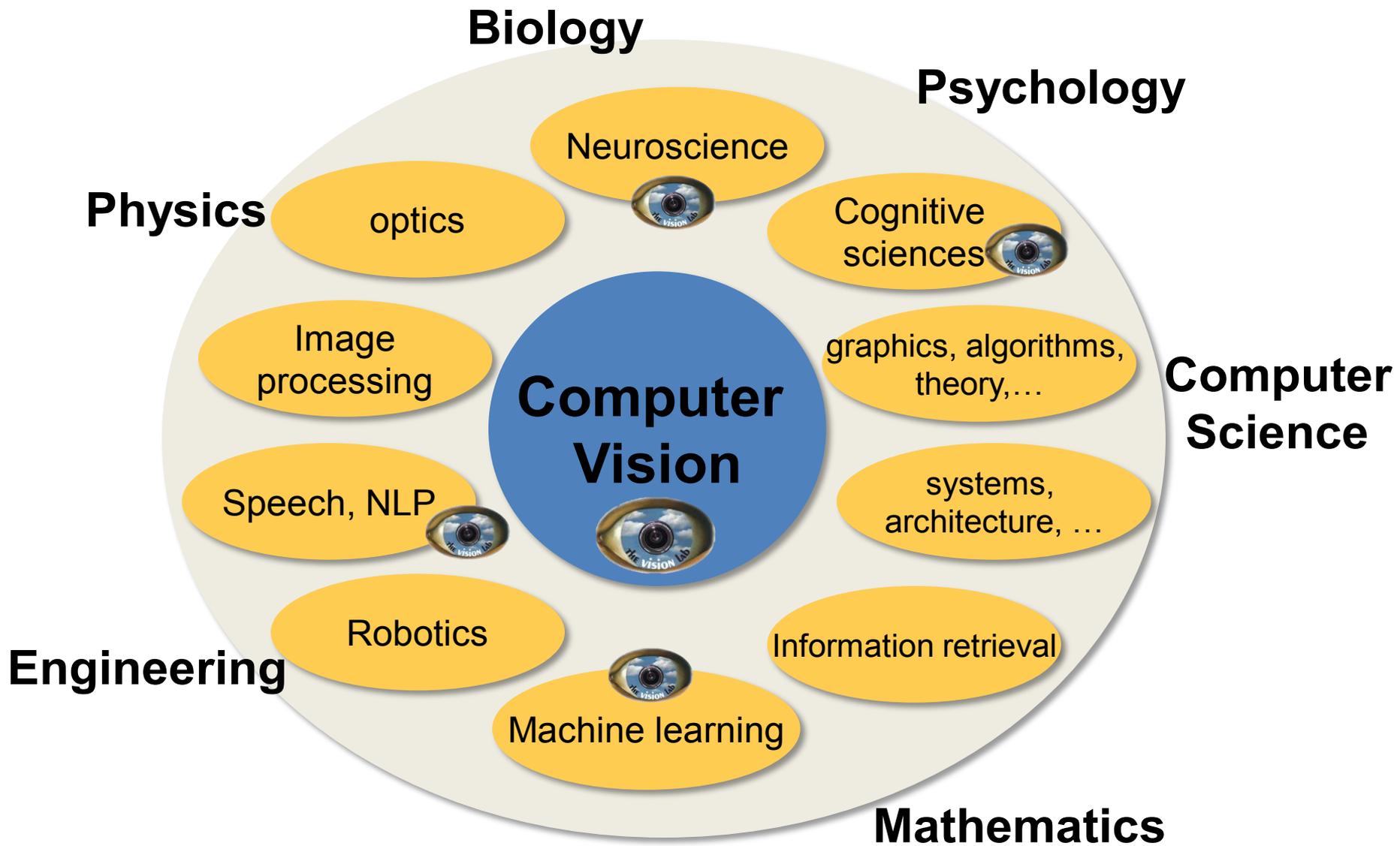


Lecture 1: Introduction

Welcome to CS231n







Computer Vision courses @ Stanford

- CS131 (fall, 2015, Profs. Fei-Fei Li & Juan Carlos Niebles):
 - Undergraduate introductory class
- CS231a (spring term, Prof. Silvio Savarese)
 - Core computer vision class for seniors, masters, and PhDs
 - Topics include image processing, cameras, 3D reconstruction, segmentation, object recognition, scene understanding
- **CS231n (this term, Prof. Fei-Fei Li & Andrej Karpathy & Justin Johnson)**
 - **Neural network (aka “deep learning”) class on image classification**
- And an assortment of CS331 and CS431 for advanced topics in computer vision

Today's agenda

- A brief history of computer vision
- CS231n overview



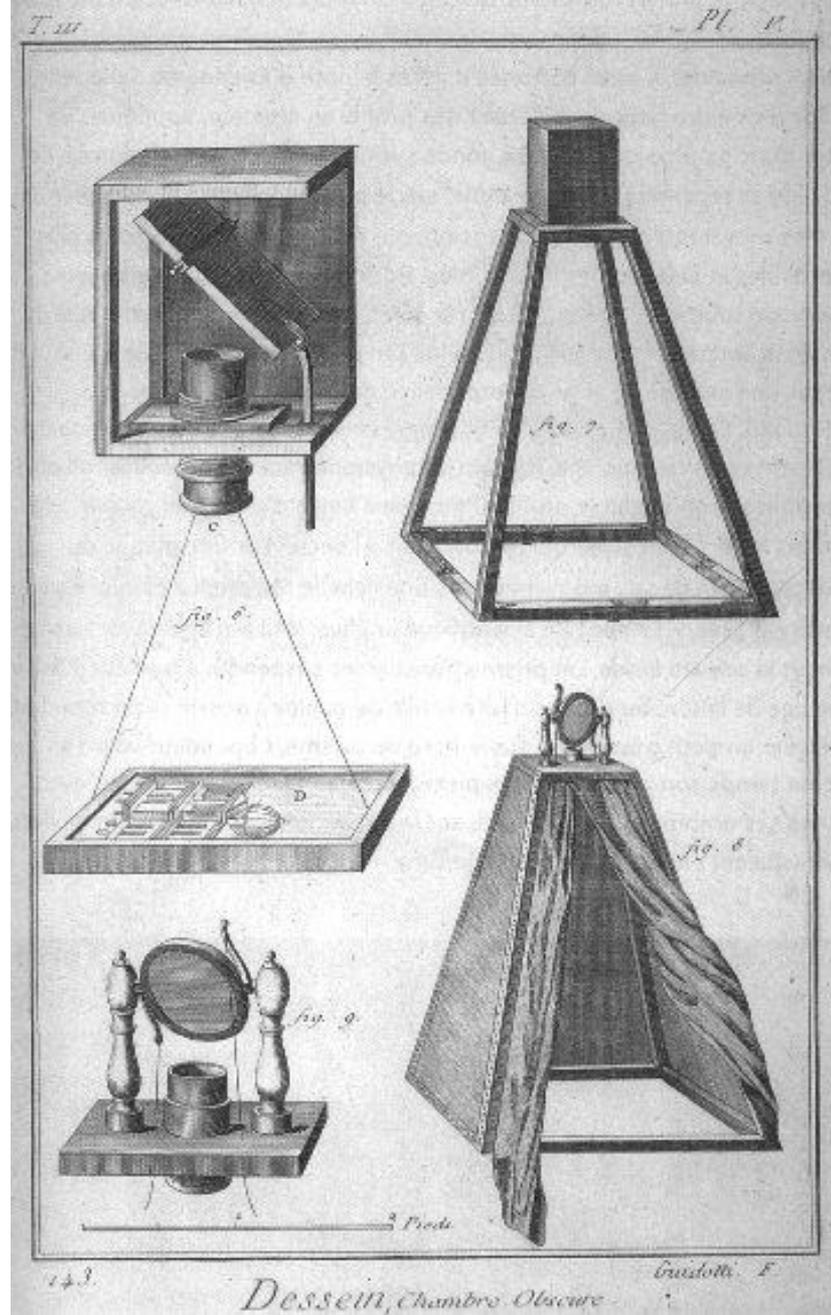
BOSNIA: THE DEAL AND THE DANGER

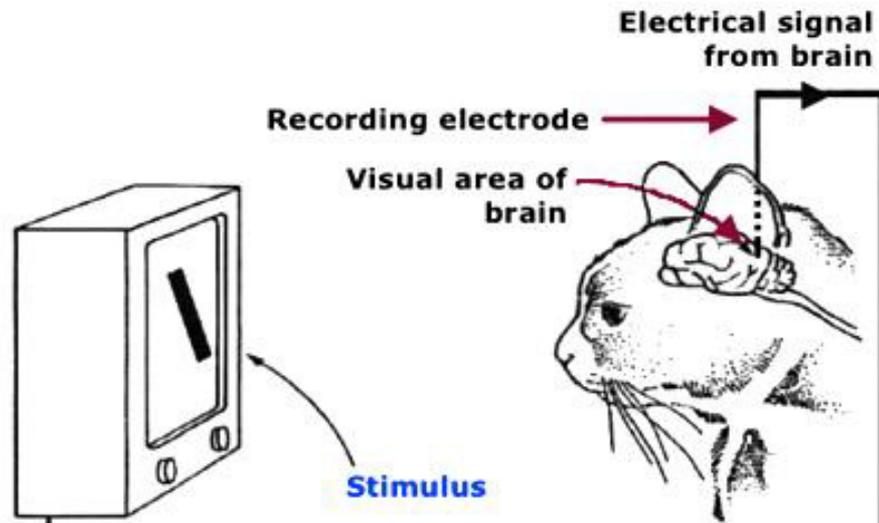
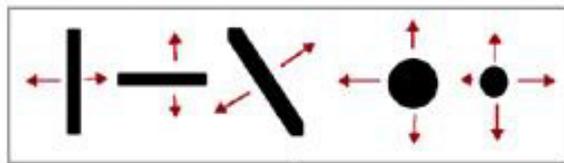


543million years, B.C.

Camera Obscura

Leonardo da Vinci
16th Century, A.D.

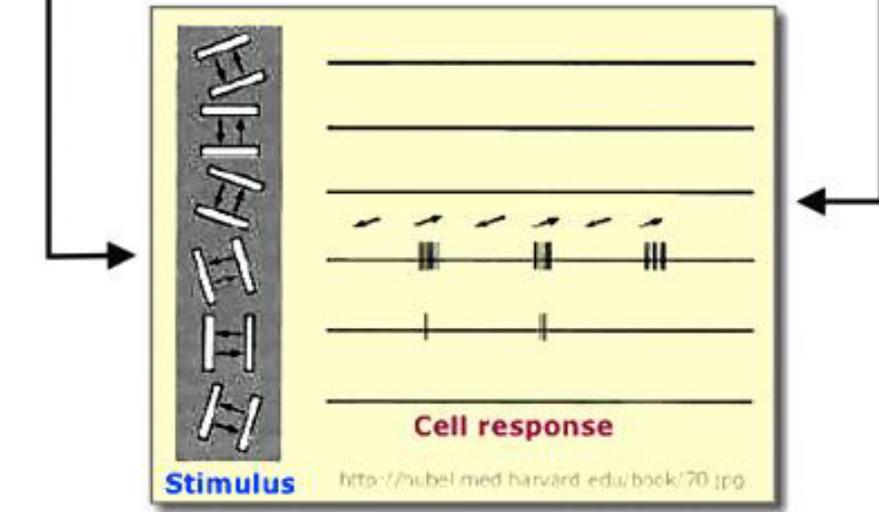
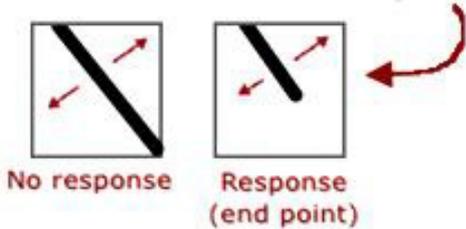




Simple Cells: Response to light orientation

Complex Cells: Response to light orientation & movement

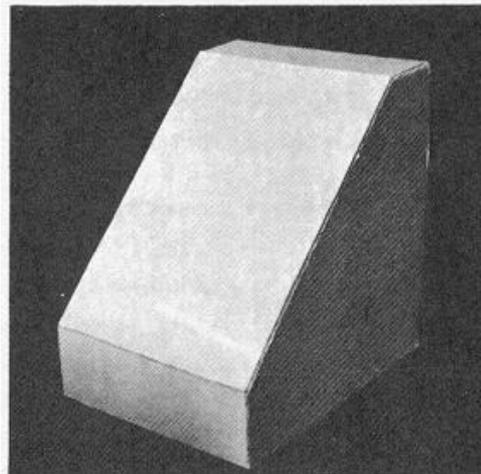
Hypercomplex Cells: Response to movement with an end point



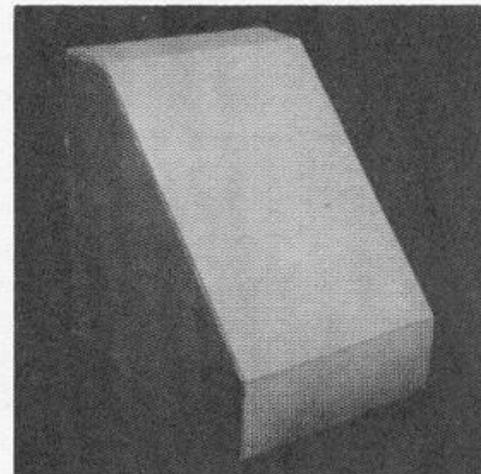
Hubel & Wiesel, 1959

Block world

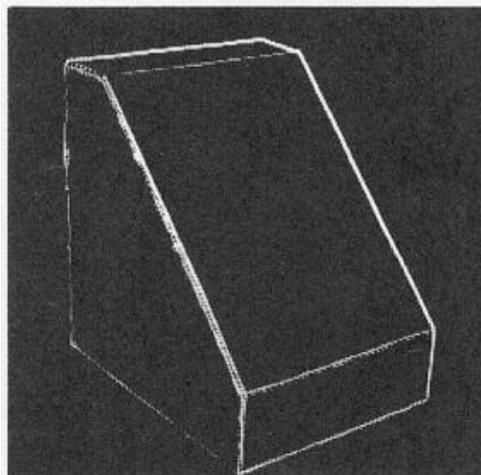
Larry Roberts,
1963



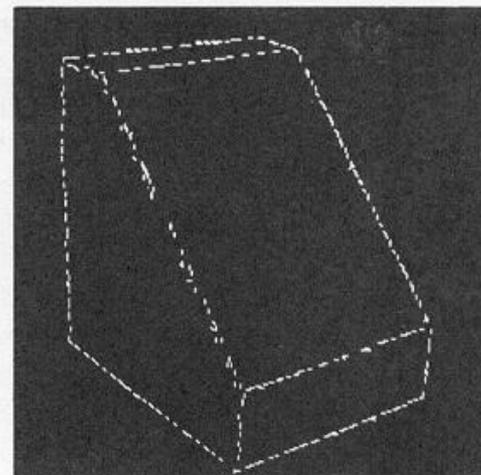
(a) Original picture.



(b) Computer display of picture
(reflected by mistake).



(c) Differentiated picture.



(d) Feature points selected.

MASSACHUSETTS INSTITUTE OF TECHNOLOGY
PROJECT MAC

Artificial Intelligence Group
Vision Memo. No. 100.

July 7, 1966

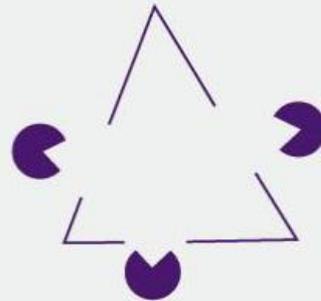
THE SUMMER VISION PROJECT

Seymour Papert

The summer vision project is an attempt to use our summer workers effectively in the construction of a significant part of a visual system. The particular task was chosen partly because it can be segmented into sub-problems which will allow individuals to work independently and yet participate in the construction of a system complex enough to be a real landmark in the development of "pattern recognition".

Copyrighted Material

VISION



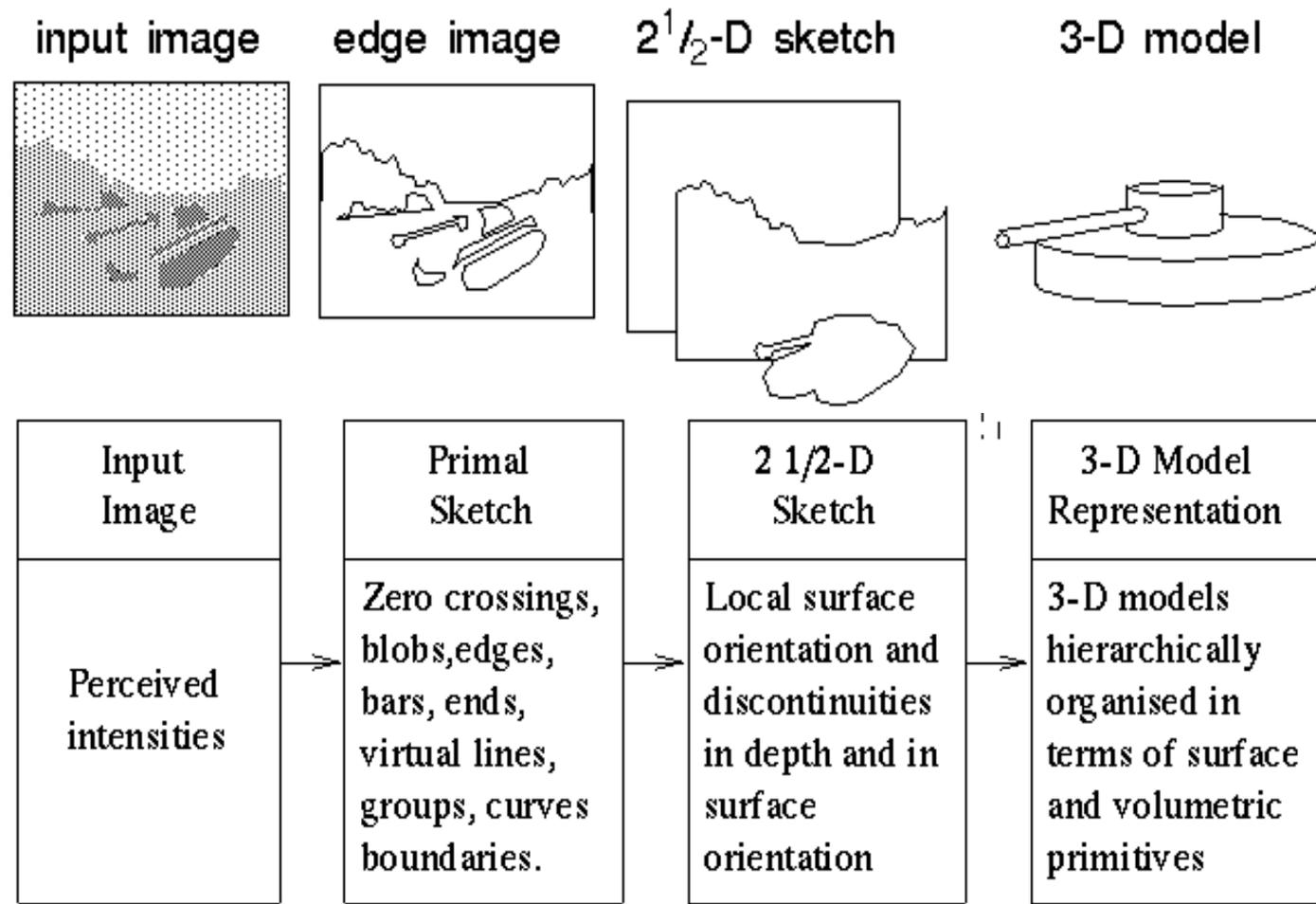
David Marr

FOREWORD BY
Shimon Ullman

AFTERWORD BY
Tomaso Poggio

Copyrighted Material

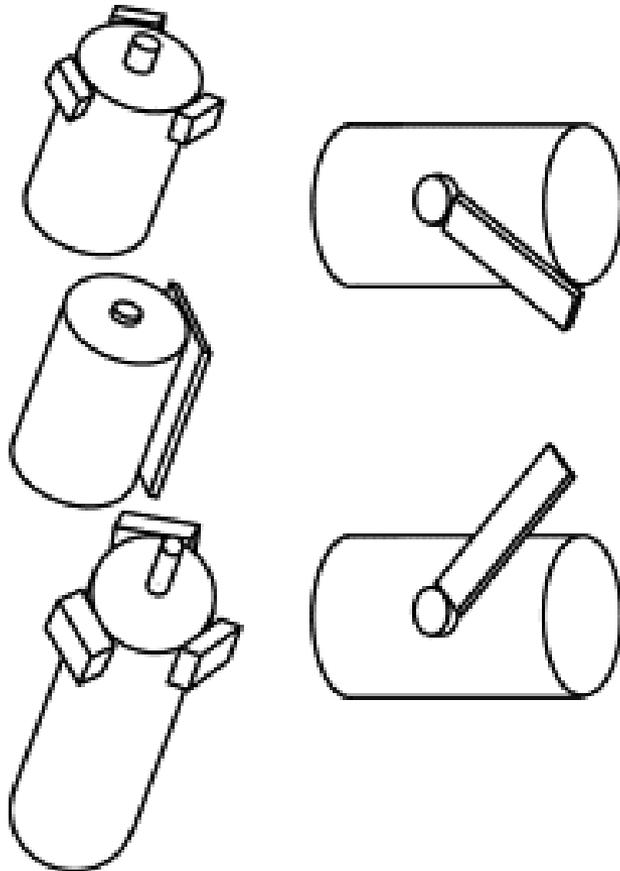
David Marr, 1970s



Stages of Visual Representation, David Marr,
1970s

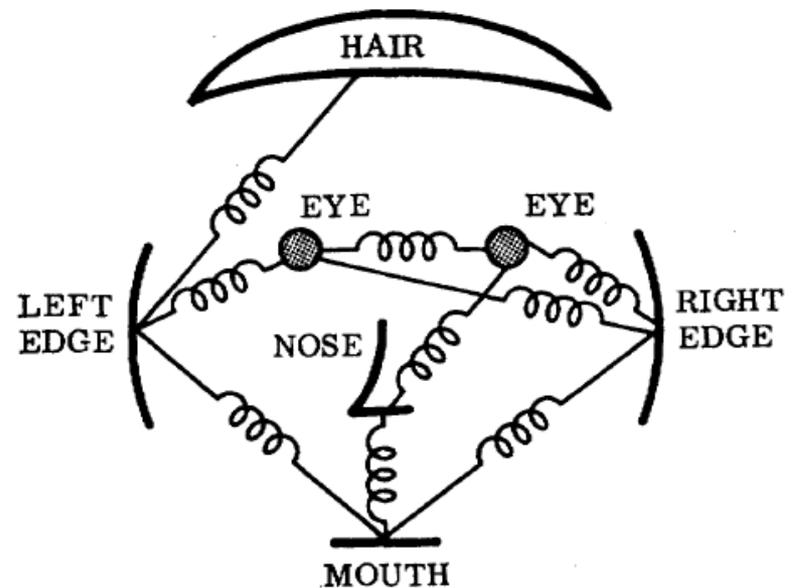
- Generalized Cylinder

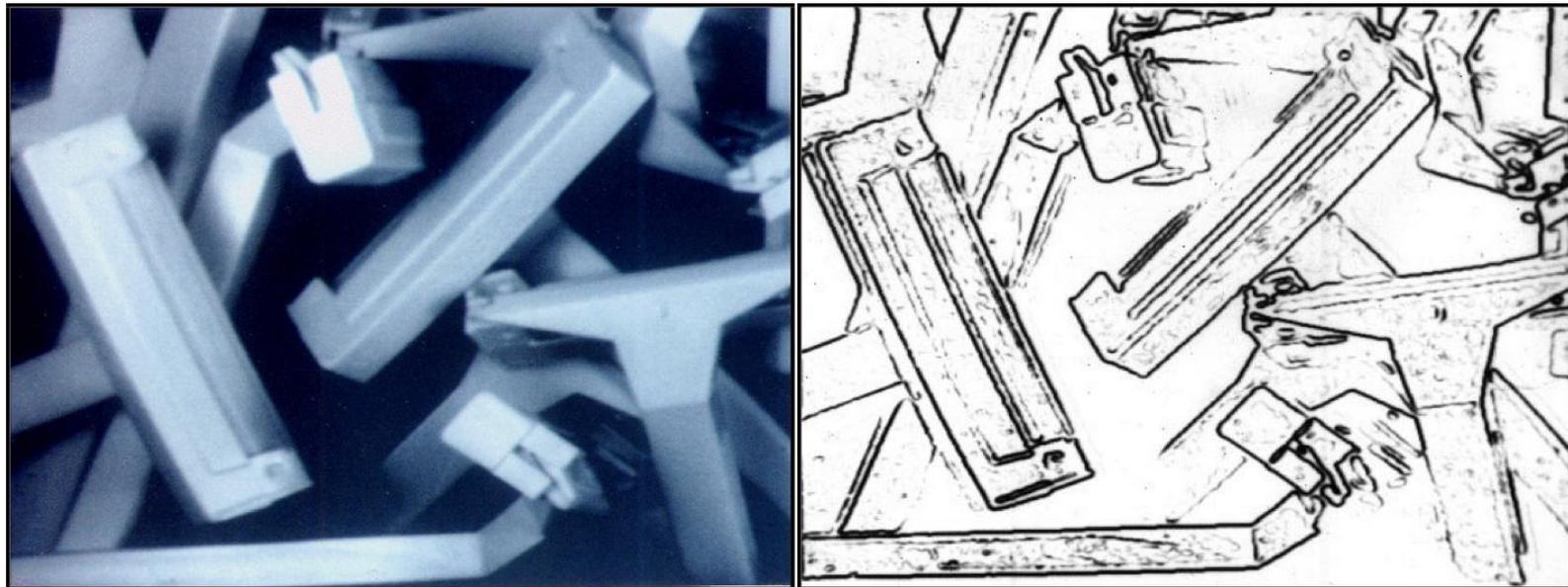
Brooks & Binford, 1979



- Pictorial Structure

Fischler and Elschlager, 1973

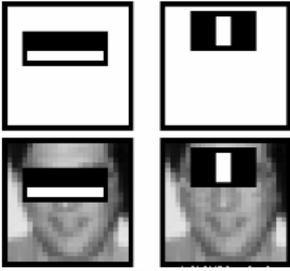




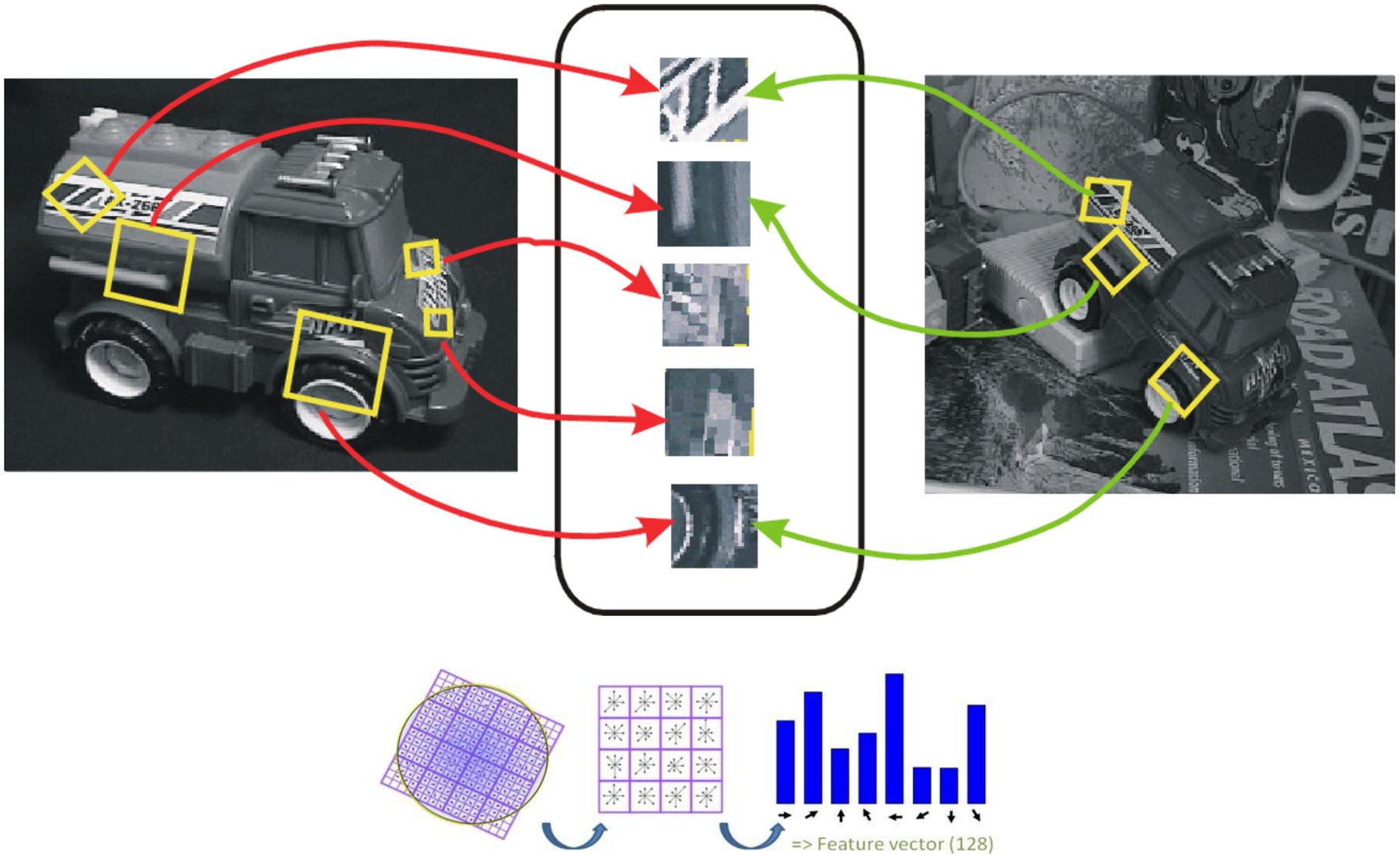
David Lowe, 1987

Normalized Cut (Shi & Malik, 1997)

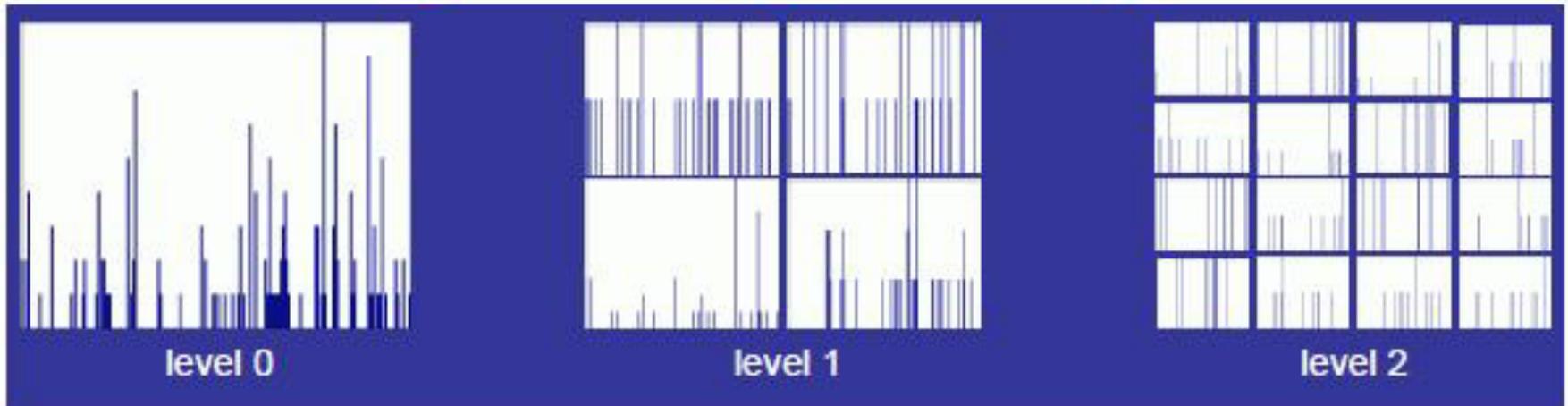
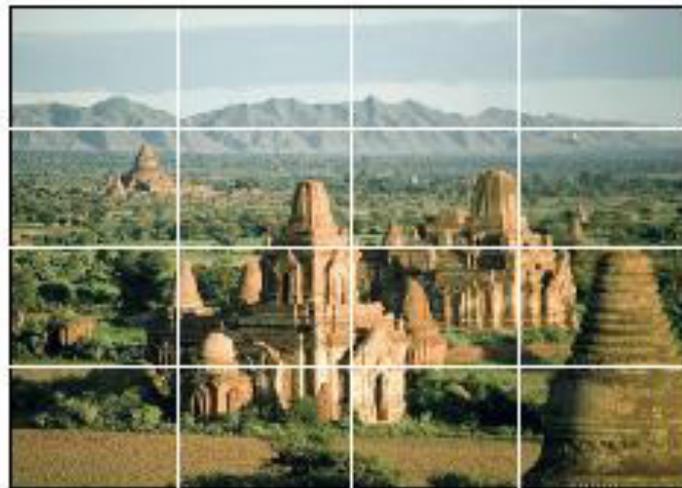




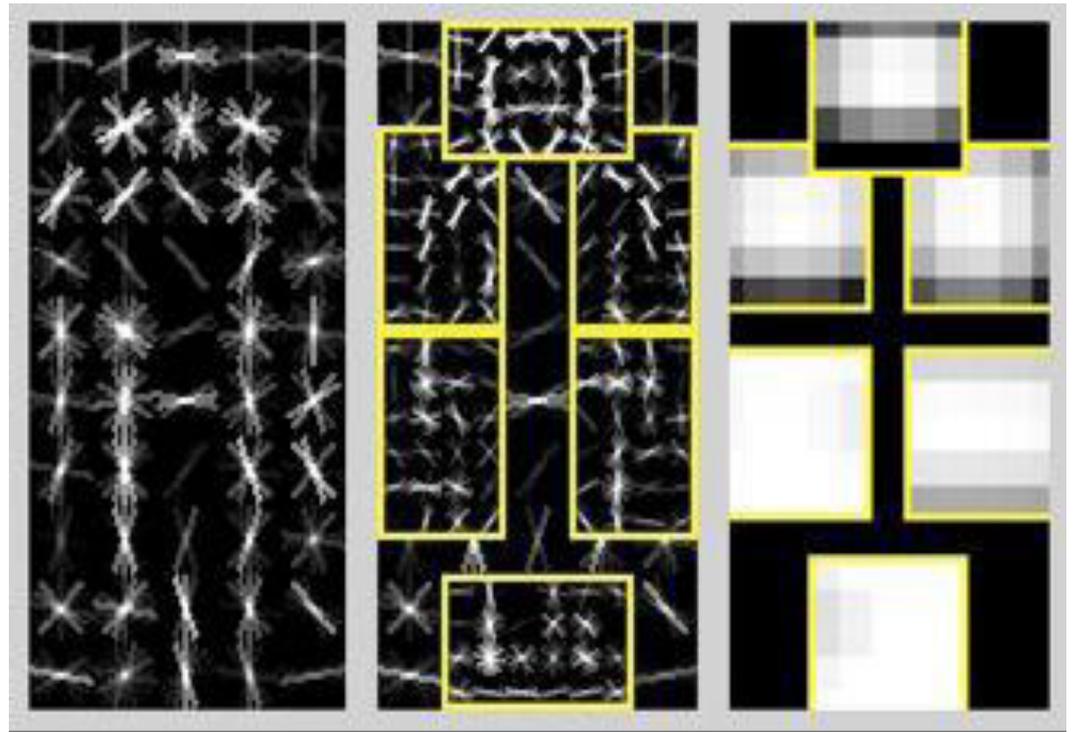
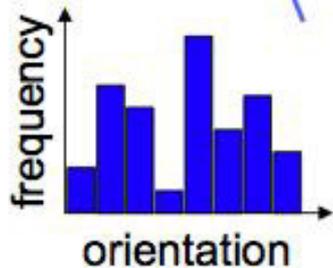
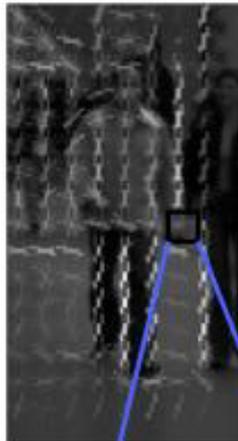
Face Detection, Viola & Jones, 2001



“SIFT” & Object Recognition, David Lowe, 1999



Spatial Pyramid Matching, Lazebnik, Schmid & Ponce, 2006

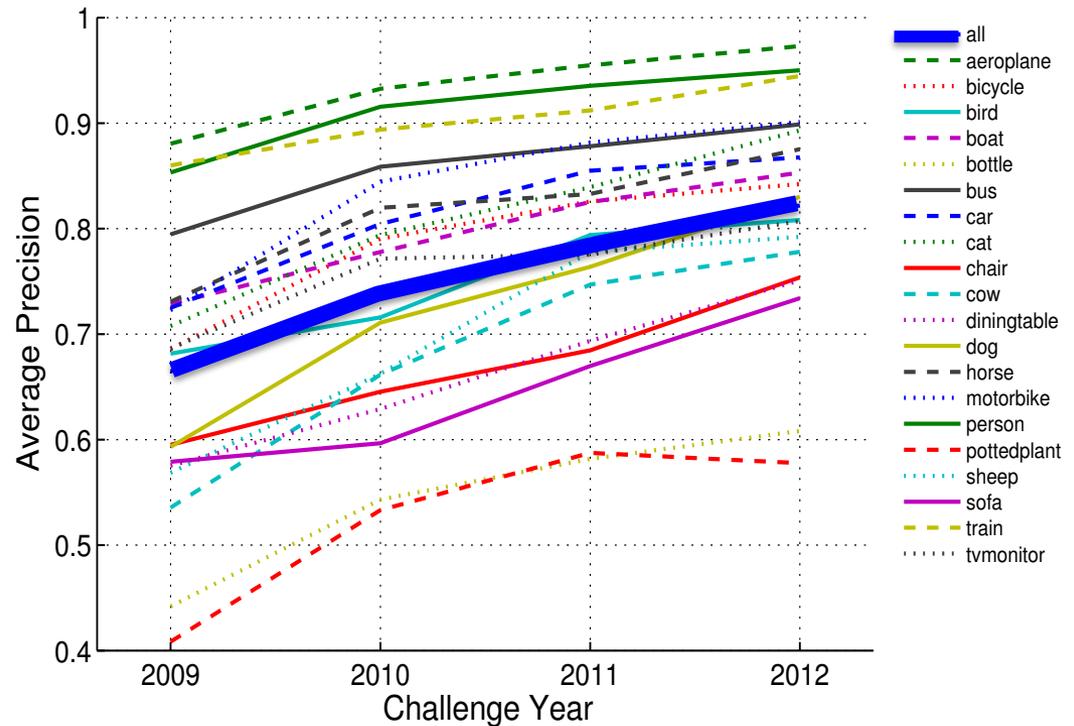


Histogram of Gradients (HoG)
Dalal & Triggs, 2005

Deformable Part Model
Felzenswalb, McAllester, Ramanan,
2009

PASCAL Visual Object Challenge (20 object categories)

[Everingham et al. 2006-2012]





www.image-net.org

22K categories and **14M** images

- Animals
 - Bird
 - Fish
 - Mammal
 - Invertebrate
- Plants
 - Tree
 - Flower
 - Food
 - Materials
- Structures
 - Artifact
 - Tools
 - Appliances
 - Structures
- Person
 - Scenes
 - Indoor
 - Geological Formations
 - Sport Activities

Deng, Dong, Socher, Li, Li, & Fei-Fei, 2009

IMAGENET Large Scale Visual Recognition Challenge

Steel drum

The Image Classification Challenge:
1,000 object classes
1,431,167 images



Output:
Scale
T-shirt
Steel drum
Drumstick
Mud turtle



Output:
Scale
T-shirt
Giant panda
Drumstick
Mud turtle

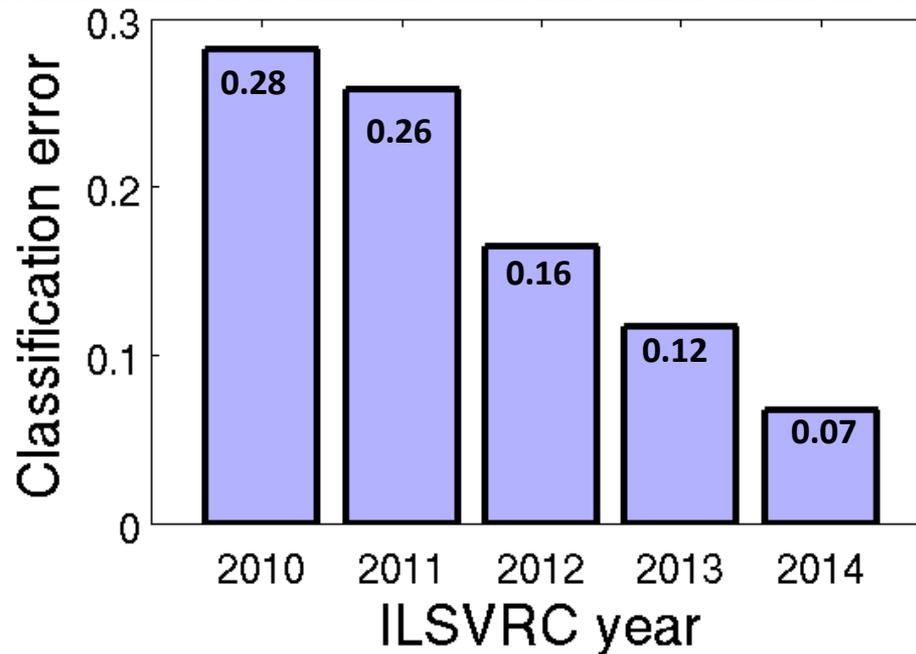


Russakovsky et al. arXiv, 2014

IMAGENET Large Scale Visual Recognition Challenge

Steel drum

The Image Classification Challenge:
1,000 object classes
1,431,167 images

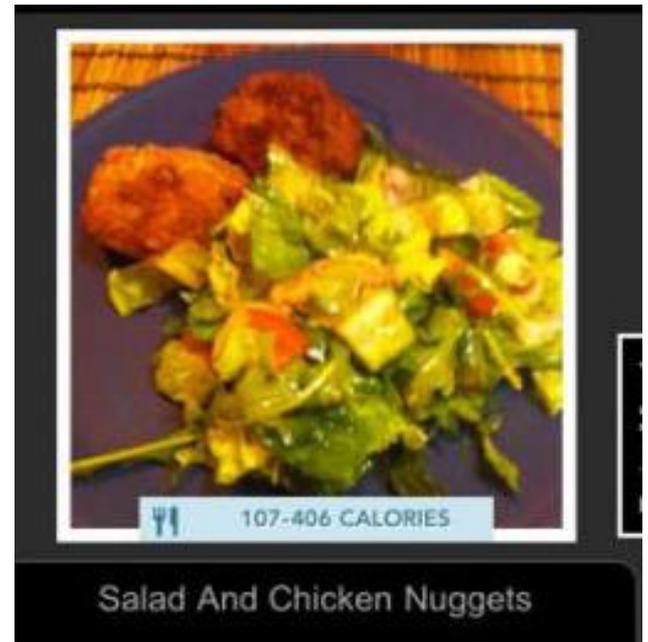


Russakovsky et al. arXiv, 2014

Today's agenda

- A brief history of computer vision
- CS231n overview

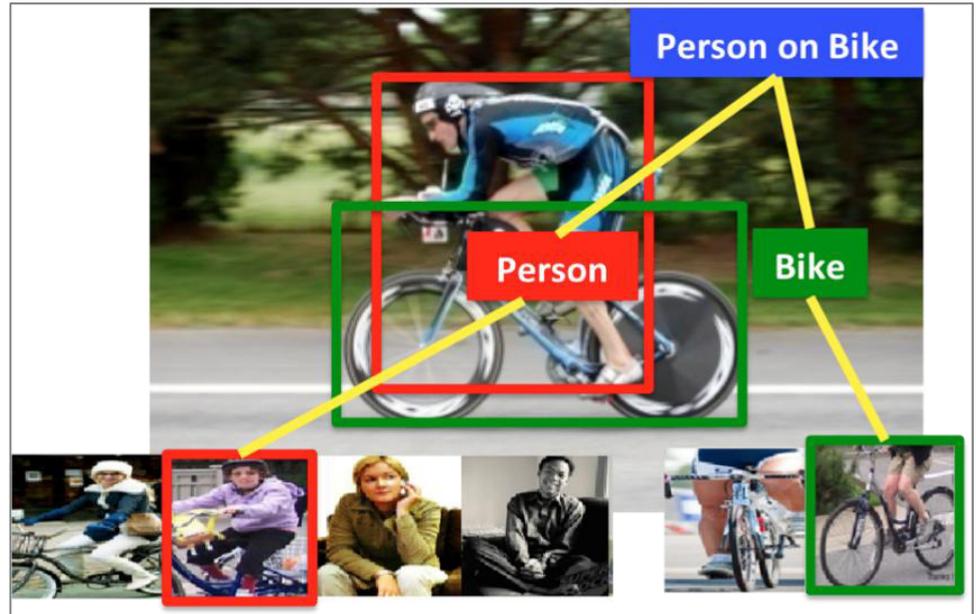
CS231n focuses on one of the most important
problems of visual recognition –
image classification



There is a number of visual recognition problems that are related to image classification, such as *object detection, image captioning*



- Object detection
- Action classification
- Image captioning
- ...



Convolutional Neural Network (CNN) has become an important tool for object recognition

IMAGENET Large Scale Visual Recognition Challenge

Year 2010

NEC-UIUC



Dense grid descriptor:
HOG, LBP

Coding: local coordinate,
super-vector

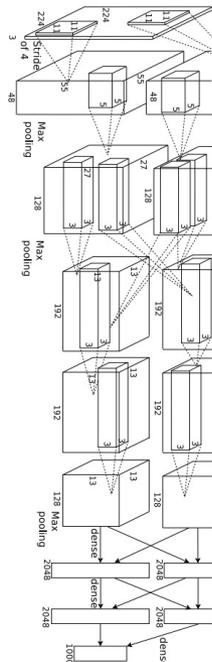
Pooling, SPM

Linear SVM

[Lin CVPR 2011]

Year 2012

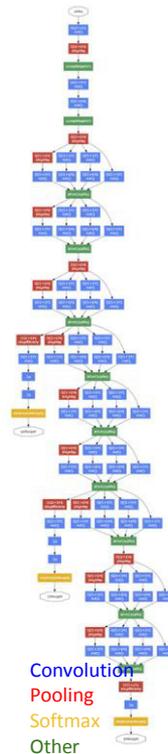
SuperVision



[Krizhevsky NIPS 2012]

Year 2014

GoogLeNet



[Szegedy arxiv 2014]

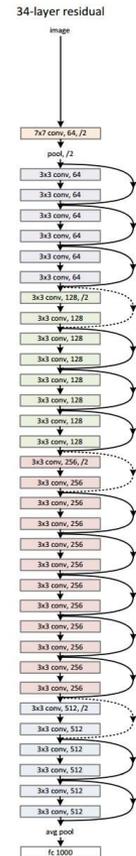
VGG



[Simonyan arxiv 2014]

Year 2015

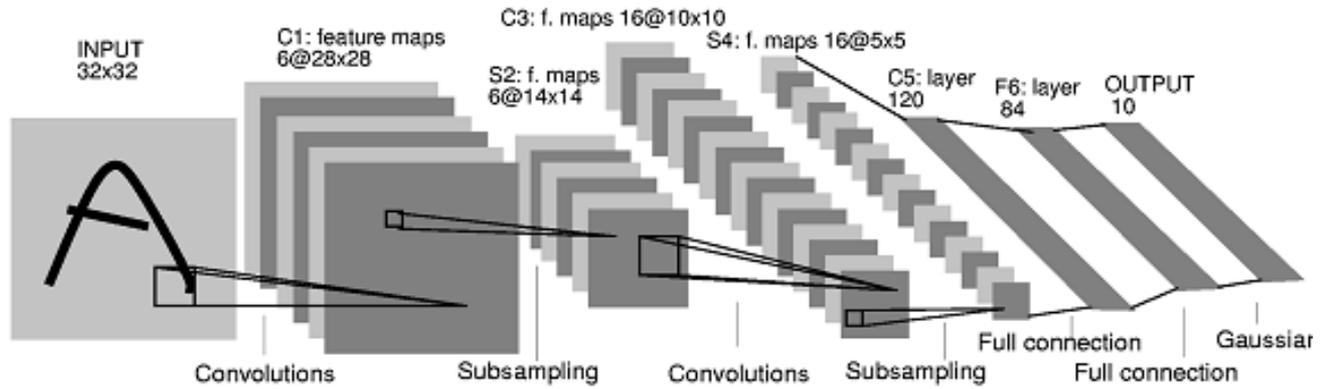
MSRA



Convolutional Neural Network (CNN)
is not invented overnight

1998

LeCun et al.



of transistors



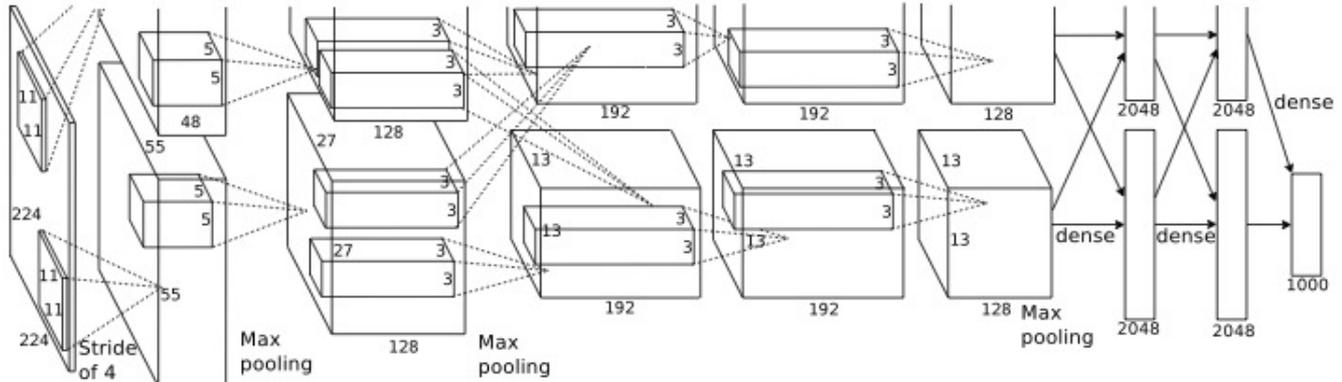
10^6

of pixels used in training

10^7 **NIST**

2012

Krizhevsky et al.



of transistors GPUs



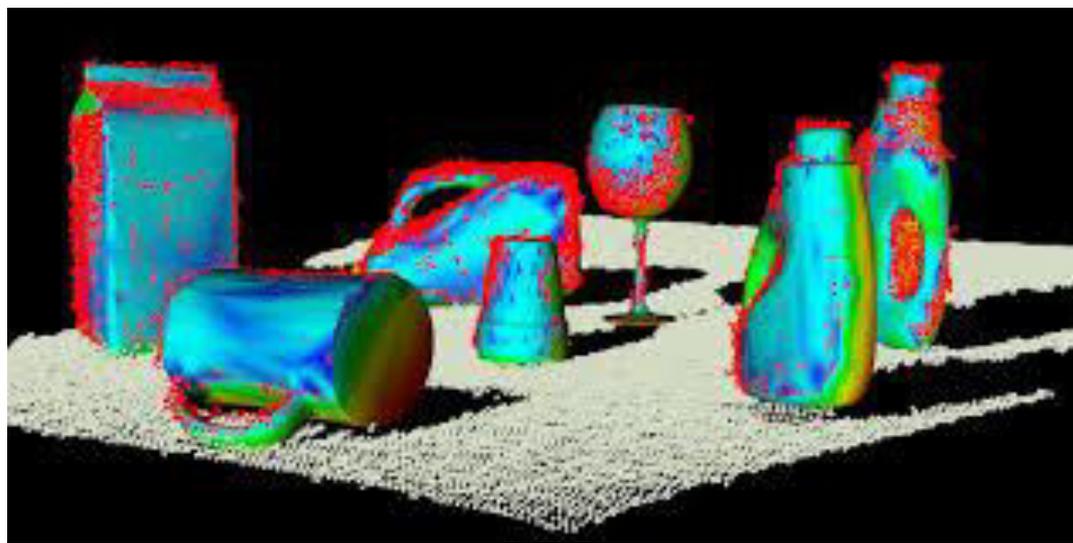
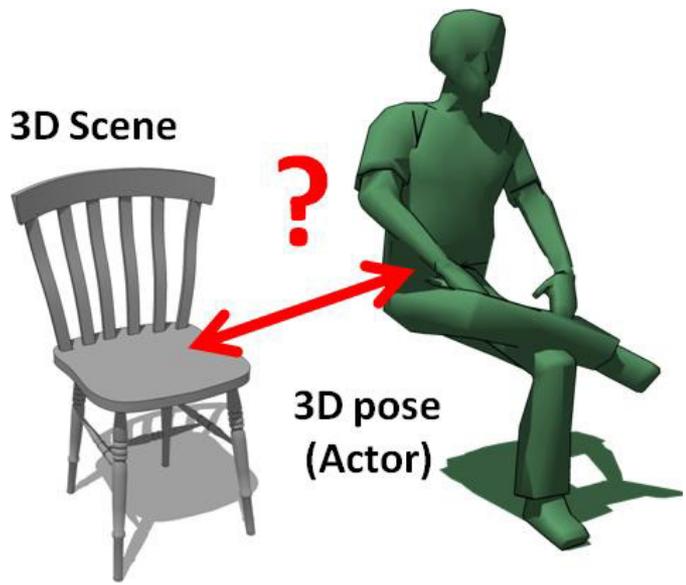
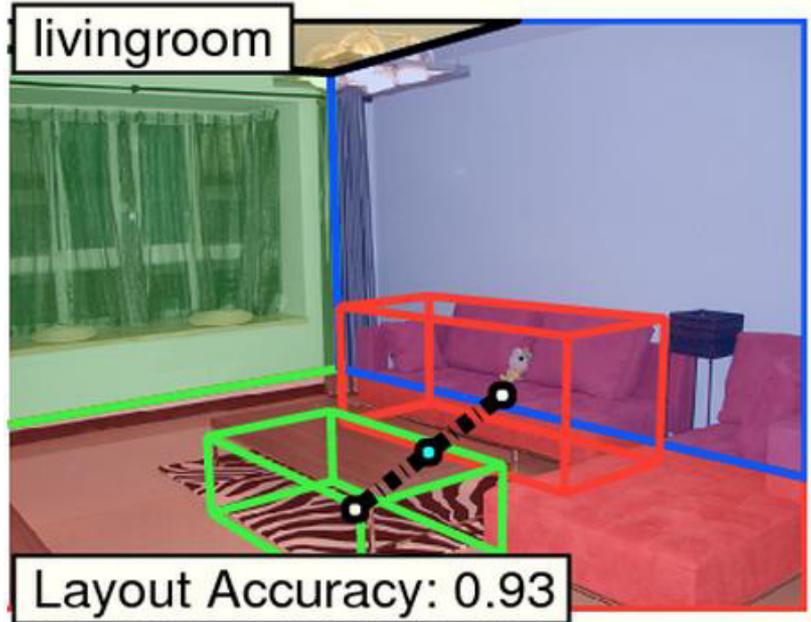
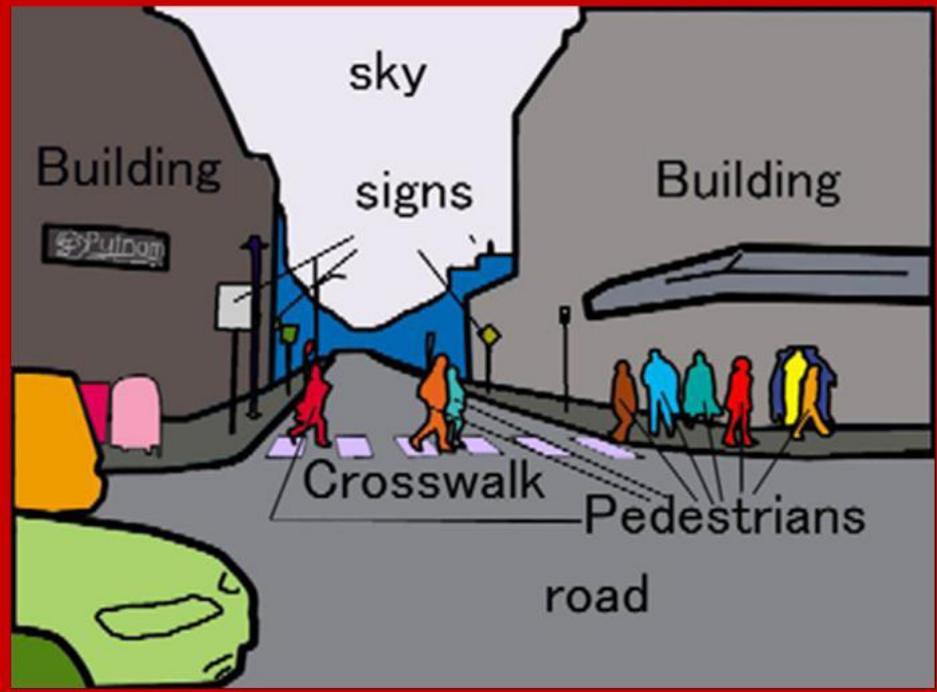
10^9

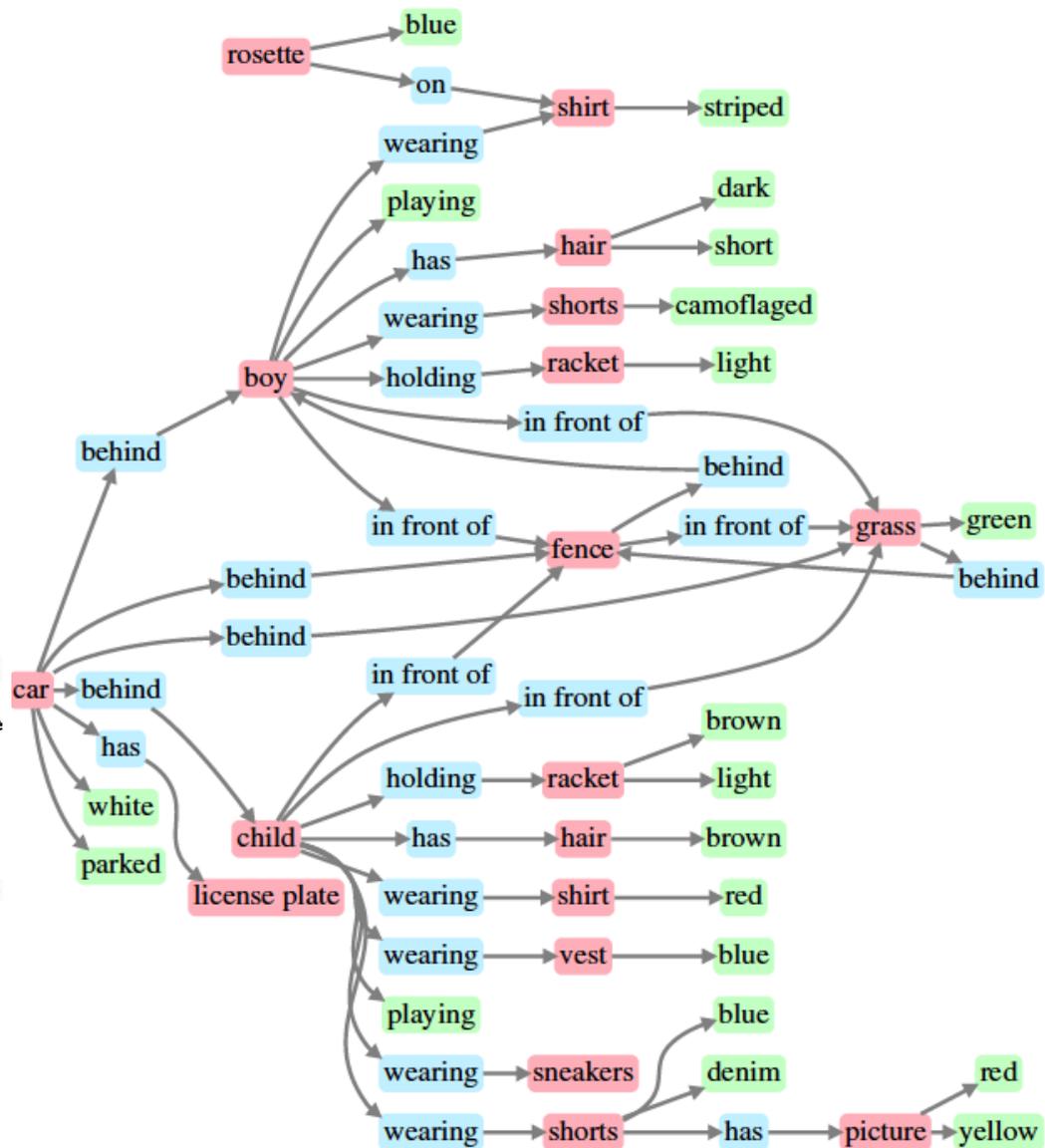
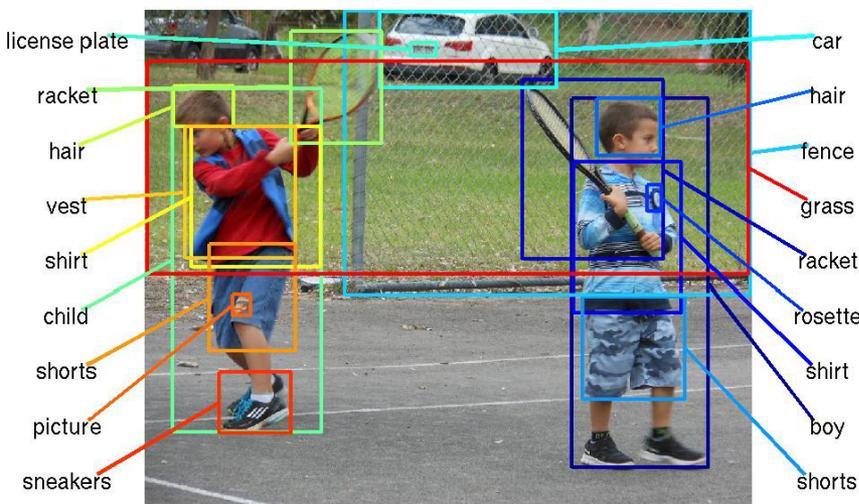


of pixels used in training

10^{14} **IMAGENET**

The quest for visual intelligence
goes far beyond object recognition...







PT = 500ms

Some kind of game or fight. Two groups of two men? The foreground pair looked like one was getting a fist in the face. Outdoors seemed like because i have an impression of grass and maybe lines on the grass? That would be why I think perhaps a game, rough game though, more like rugby than football because they pairs weren't in pads and helmets, though I did get the impression of similar clothing. maybe some trees? in the background. (Subject: SM)

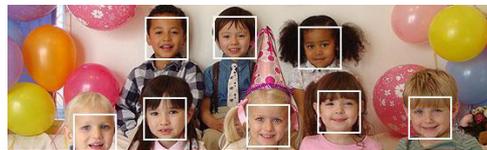
Fei-Fei, Iyer, Koch, Perona, *JoV*, 2007



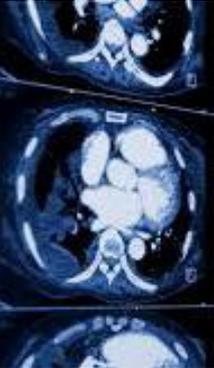
The picture above is funny.

But for me it is also one of those examples that make me sad about the outlook for AI and for Computer Vision. What would it take for a computer to understand this image as you or I do? I challenge you to think explicitly of all the pieces of knowledge that have to fall in place for it to make sense. Here is my short attempt:

- You recognize it is an image of a bunch of people and you understand they are in a hallway
- You recognize that there are 3 mirrors in the scene so some of those people are "fake" replicas from different viewpoints.
- You recognize Obama from the few pixels that make up his face. It helps that he is in his suit and that he is surrounded by other people with suits.
- You recognize that there's a person standing on a scale, even though the scale occupies only very few white pixels that blend with the background. But, you've used the person's pose and knowledge of how people interact with objects to figure it out.
- You recognize that Obama has his foot positioned just slightly on top of the scale. Notice the language I'm using: It is in terms of the 3D structure of the scene, not the position of the leg in the 2D coordinate system of the image.
- You know how physics works: Obama is leaning in on the scale, which applies a force on it. Scale measures force that is applied on it, that's how it works => it will over-estimate the weight of the person standing on it.
- The person measuring his weight is not aware of Obama doing this. You derive this because you know his pose, you understand that the field of view of a person is finite, and you understand that he is not very likely to sense the slight push of Obama's foot.
- You understand that people are self-conscious about their weight. You also understand that he is reading off the scale measurement, and that shortly the over-estimated weight will confuse him because it will probably be much higher than what he expects. In other words, you reason about implications of the events that are about to unfold seconds after this photo was taken, and especially about the thoughts and how they will develop inside people's heads. You also reason about what pieces of information are available to people.
- There are people in the back who find the person's imminent confusion funny. In other words you are reasoning about state of mind of people, and their view of the state of mind of another person. That's getting frighteningly meta.
- Finally, the fact that the perpetrator here is the president makes it maybe even a little more funnier. You understand what actions are more or less likely to be undertaken by different people based on their status and identity.



Computer Vision Technology Can Better Our Lives



Who we are

- Instructors



Fei-Fei Li



Andrej Karpathy



Justin Johnson

- Teaching Assistants



Serena Yeung



Subhasis Das



Song Han



Albert Haque



Bharath Ramsundar



Hieu Pham



Irawn Bello

- Keeping in touch:

- cs231n-winter1516-staff@lists.stanford.edu

- Piazza

Our philosophy

- Thorough and Detailed.
 - Understand how to write from scratch, debug and train convolutional neural networks.
- Practical.
 - Focus on practical techniques for training these networks at scale, and on GPUs (e.g. will touch on distributed optimization, differences between CPU vs. GPU, etc.) Also look at state of the art software tools such as Caffe, maybe also Torch and TensorFlow
- State of the art.
 - Most materials are new from research world in the past 1-3 years. Very exciting stuff!
- Fun.
 - Some fun topics such as Image Captioning (using RNN)
 - Also DeepDream, NeuralStyle, etc.

Our philosophy (cont'd)

- Fun.
 - Some fun topics such as Image Captioning (using RNN)
 - Also DeepDream, NeuralStyle, etc.



Grading policy

- 3 Problem Sets: $15\% \times 3 = 45\%$
- Midterm Exam: 15%
- Final Course Project: 40%
 - Milestone: 5%
 - Final write-up: 35%
 - Bonus points for exceptional poster presentation
- Late policy
 - 7 free late days – use them in your ways
 - Afterwards, 25% off per day late
 - Not accepted after 3 late days per PS
 - Does not apply to Final Course Project
- Collaboration policy
 - Read the student code book, understand what is ‘collaboration’ and what is ‘academic infraction’

Pre-requisite

- Proficiency in Python, some high-level familiarity with C/C++
 - All class assignments will be in Python (and use numpy), but some of the deep learning libraries we may look at later in the class are written in C++.
 - A Python tutorial available on course website
- College Calculus, Linear Algebra
- Equivalent knowledge of CS229 (Machine Learning)
 - We will be formulating cost functions, taking derivatives and performing optimization with gradient descent.

Syllabus

- Go to website...

<http://vision.stanford.edu/teaching/cs231n/index.html>

References

- Hubel, David H., and Torsten N. Wiesel. "Receptive fields, binocular interaction and functional architecture in the cat's visual cortex." *The Journal of physiology* 160.1 (1962): 106. [\[PDF\]](#)
- Roberts, Lawrence Gilman. "Machine Perception of Three-dimensional Solids." Diss. Massachusetts Institute of Technology, 1963. [\[PDF\]](#)
- Marr, David. "Vision." The MIT Press, 1982. [\[PDF\]](#)
- Brooks, Rodney A., and Creiner, Russell and Binford, Thomas O. "The ACRONYM model-based vision system. " In *Proceedings of the 6th International Joint Conference on Artificial Intelligence* (1979): 105-113. [\[PDF\]](#)
- Fischler, Martin A., and Robert A. Elschlager. "The representation and matching of pictorial structures." *IEEE Transactions on Computers* 22.1 (1973): 67-92. [\[PDF\]](#)
- Lowe, David G., "Three-dimensional object recognition from single two-dimensional images," *Artificial Intelligence*, 31, 3 (1987), pp. 355-395. [\[PDF\]](#)
- Shi, Jianbo, and Jitendra Malik. "Normalized cuts and image segmentation." *Pattern Analysis and Machine Intelligence*, *IEEE Transactions on* 22.8 (2000): 888-905. [\[PDF\]](#)
- Viola, Paul, and Michael Jones. "Rapid object detection using a boosted cascade of simple features." *Computer Vision and Pattern Recognition*, 2001. *CVPR 2001. Proceedings of the 2001 IEEE Computer Society Conference on*. Vol. 1. IEEE, 2001. [\[PDF\]](#)
- Lowe, David G. "Distinctive image features from scale-invariant keypoints." *International Journal of Computer Vision* 60.2 (2004): 91-110. [\[PDF\]](#)
- Lazebnik, Svetlana, Cordelia Schmid, and Jean Ponce. "Beyond bags of features: Spatial pyramid matching for recognizing natural scene categories." *Computer Vision and Pattern Recognition*, 2006 *IEEE Computer Society Conference on*. Vol. 2. IEEE, 2006. [\[PDF\]](#)

- Dalal, Navneet, and Bill Triggs. "Histograms of oriented gradients for human detection." Computer Vision and Pattern Recognition, 2005. CVPR 2005. IEEE Computer Society Conference on. Vol. 1. IEEE, 2005. [\[PDF\]](#)
- Felzenszwalb, Pedro, David McAllester, and Deva Ramanan. "A discriminatively trained, multiscale, deformable part model." Computer Vision and Pattern Recognition, 2008. CVPR 2008. IEEE Conference on. IEEE, 2008 [\[PDF\]](#)
- Everingham, Mark, et al. "The pascal visual object classes (VOC) challenge." International Journal of Computer Vision 88.2 (2010): 303-338. [\[PDF\]](#)
- Deng, Jia, et al. "Imagenet: A large-scale hierarchical image database." Computer Vision and Pattern Recognition, 2009. CVPR 2009. IEEE Conference on. IEEE, 2009. [\[PDF\]](#)
- Russakovsky, Olga, et al. "Imagenet Large Scale Visual Recognition Challenge." arXiv:1409.0575. [\[PDF\]](#)
- Lin, Yuanqing, et al. "Large-scale image classification: fast feature extraction and SVM training." Computer Vision and Pattern Recognition (CVPR), 2011 IEEE Conference on. IEEE, 2011. [\[PDF\]](#)
- Krizhevsky, Alex, Ilya Sutskever, and Geoffrey E. Hinton. "Imagenet classification with deep convolutional neural networks." Advances in neural information processing systems. 2012. [\[PDF\]](#)
- Szegedy, Christian, et al. "Going deeper with convolutions." arXiv preprint arXiv:1409.4842 (2014). [\[PDF\]](#)
- Simonyan, Karen, and Andrew Zisserman. "Very deep convolutional networks for large-scale image recognition." arXiv preprint arXiv:1409.1556 (2014). [\[PDF\]](#)
- He, Kaiming, et al. "Spatial Pyramid Pooling in Deep Convolutional Networks for Visual Recognition." arXiv preprint arXiv:1406.4729 (2014). [\[PDF\]](#)
- LeCun, Yann, et al. "Gradient-based learning applied to document recognition." Proceedings of the IEEE 86.11 (1998): 2278-2324. [\[PDF\]](#)
- Fei-Fei, Li, et al. "What do we perceive in a glance of a real-world scene?." Journal of vision 7.1 (2007): 10. [\[PDF\]](#)