

CS543 / ECE549
Computer Vision
Spring 2020

Course webpage URL: <https://s-gupta.github.io/ece549/>

The goal of computer vision

- To extract “meaning” from pixels



What we see

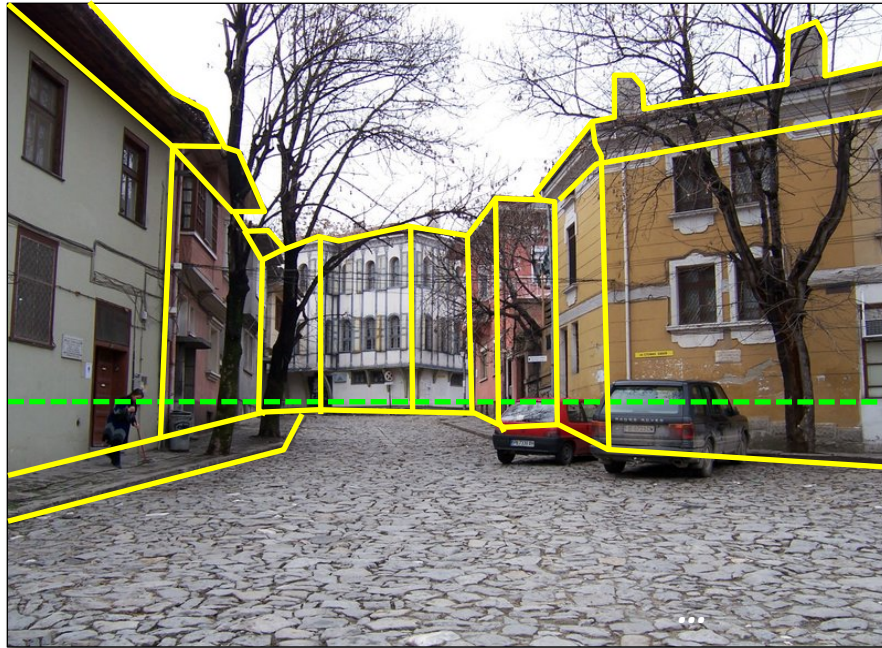
0	3	2	5	4	7	6	9	8
3	0	1	2	3	4	5	6	7
2	1	0	3	2	5	4	7	6
5	2	3	0	1	2	3	4	5
4	3	2	1	0	3	2	5	4
7	4	5	2	3	0	1	2	3
6	5	4	3	2	1	0	3	2
9	6	7	4	5	2	3	0	1
8	7	6	5	4	3	2	1	0

What a computer sees

What kind of information can be extracted from an image?

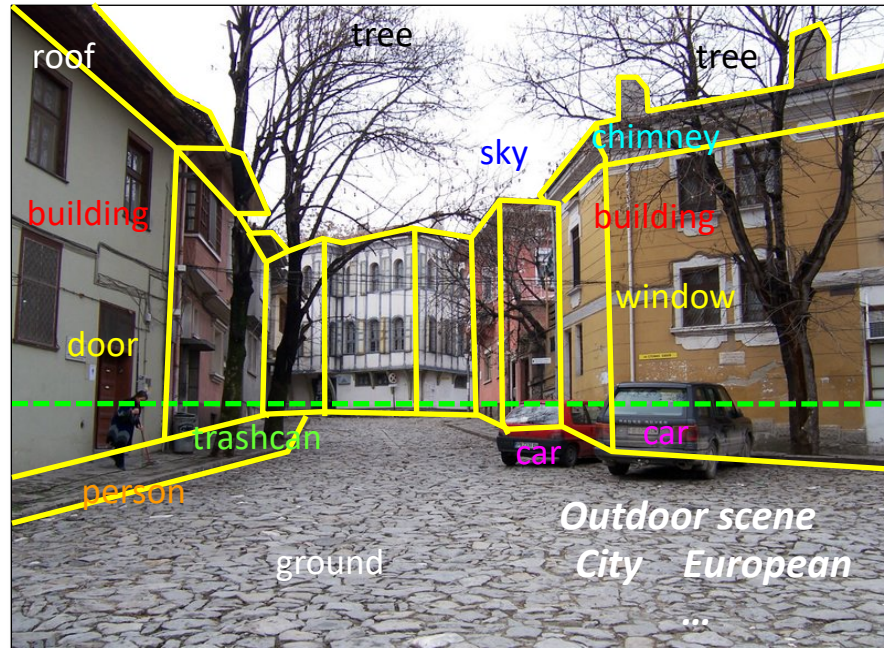


What kind of information can be extracted from an image?



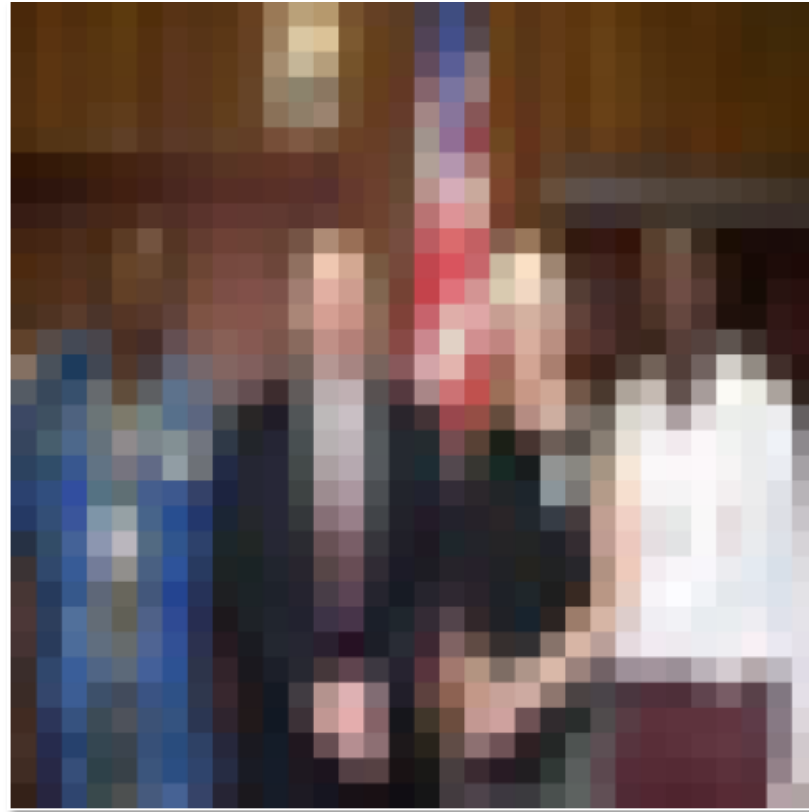
Geometric information

What kind of information can be extracted from an image?



Geometric information
Semantic information

Vision is easy for humans



Source: L. Lazebnik

Source: "80 million tiny images" by Torralba et al.

Vision is easy for humans

Attneave's Cat

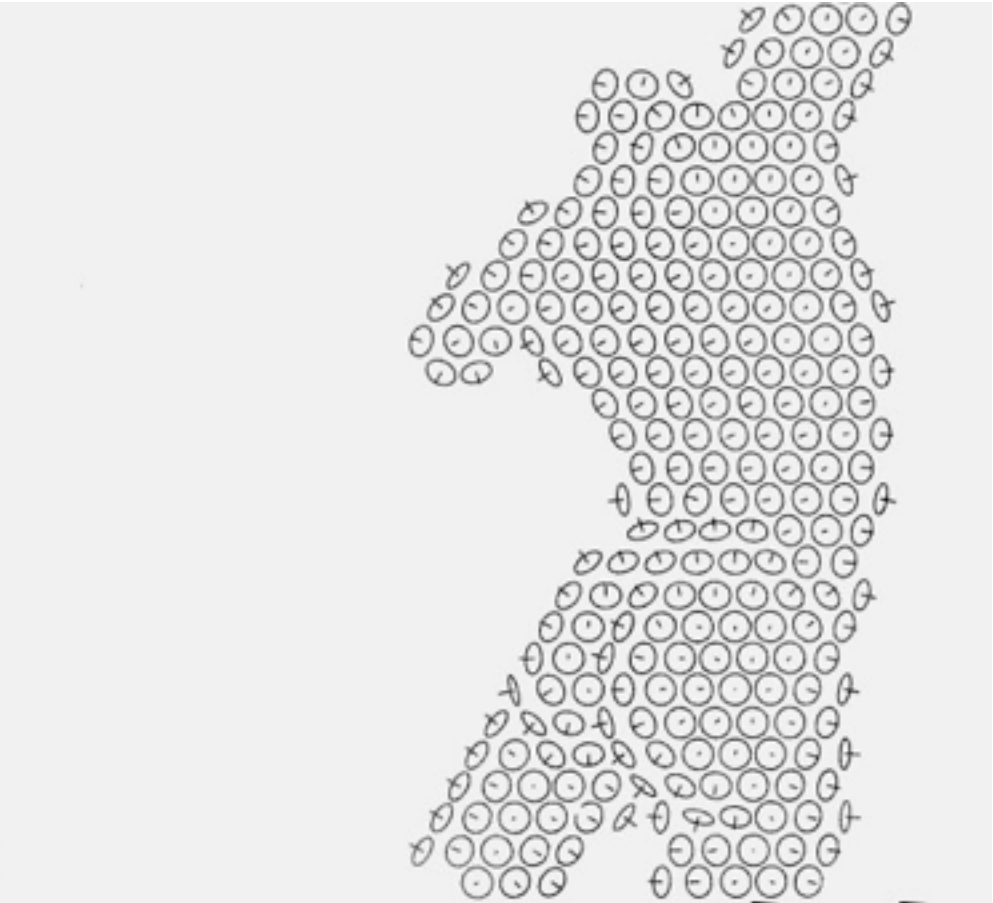
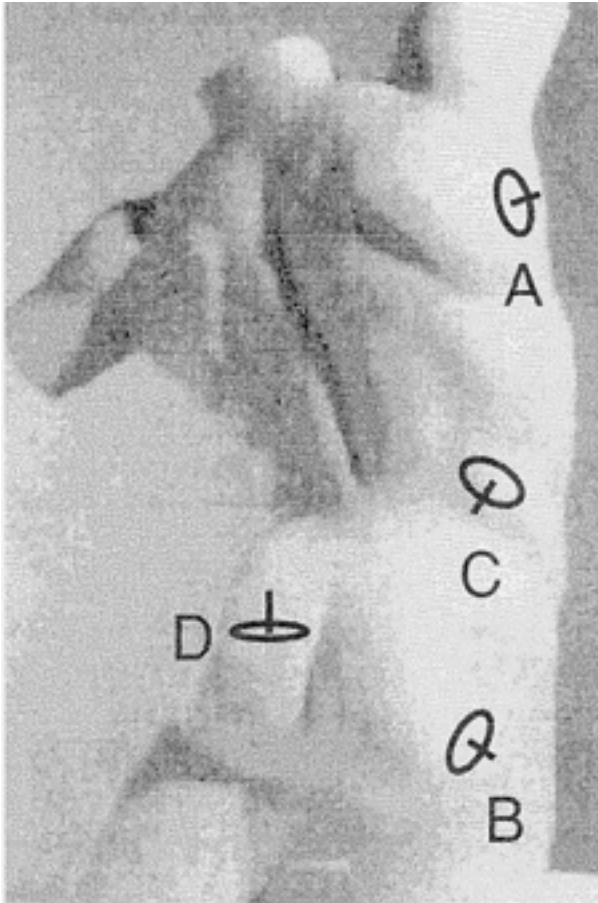


Vision is easy for humans

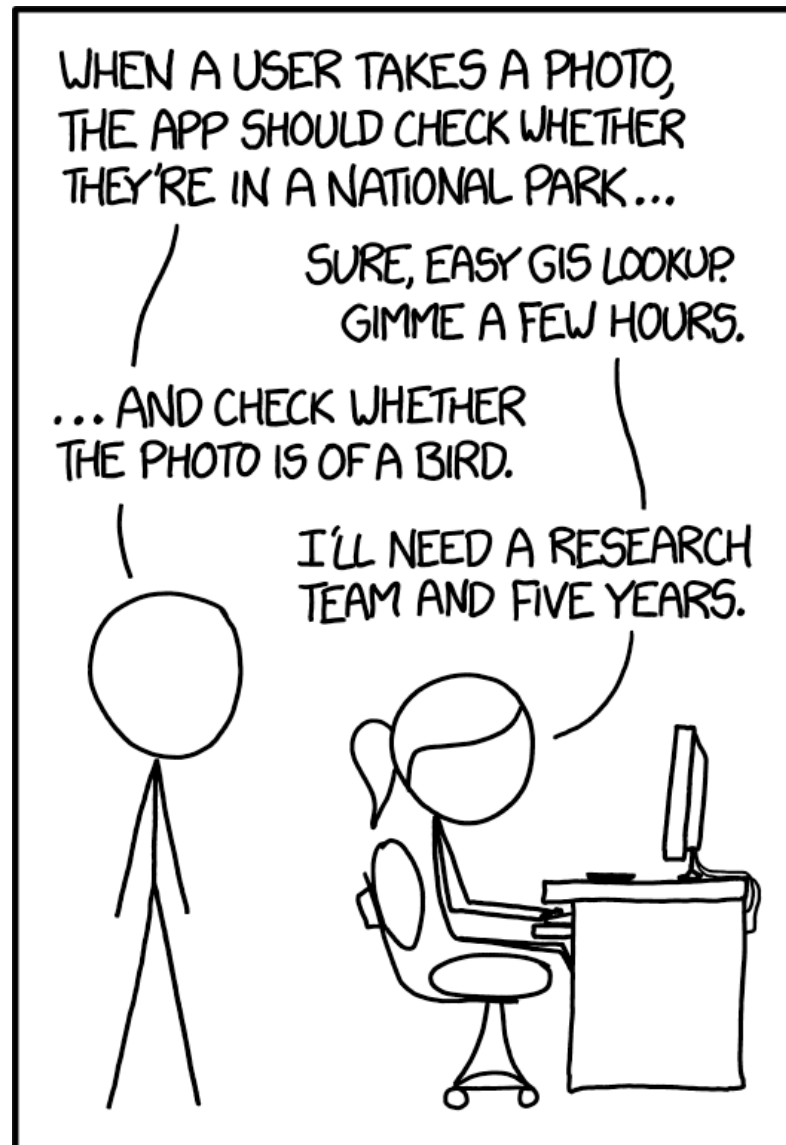
Mooney Faces



Vision is easy for humans



Remarkably Hard for Computers



Vision is hard: Images are ambiguous



Vision is hard: Objects Blend Together



the superficial

Vision is hard: Objects Blend Together



Vision is hard: Intra-class Variation



Viewpoint variation



Illumination



Scale

Vision is hard: Intra-class Variation



Shape variation



Occlusion



Background clutter

Vision is hard: Intra-class Variation



Vision is hard: Concepts are subtle



Tennessee Warbler

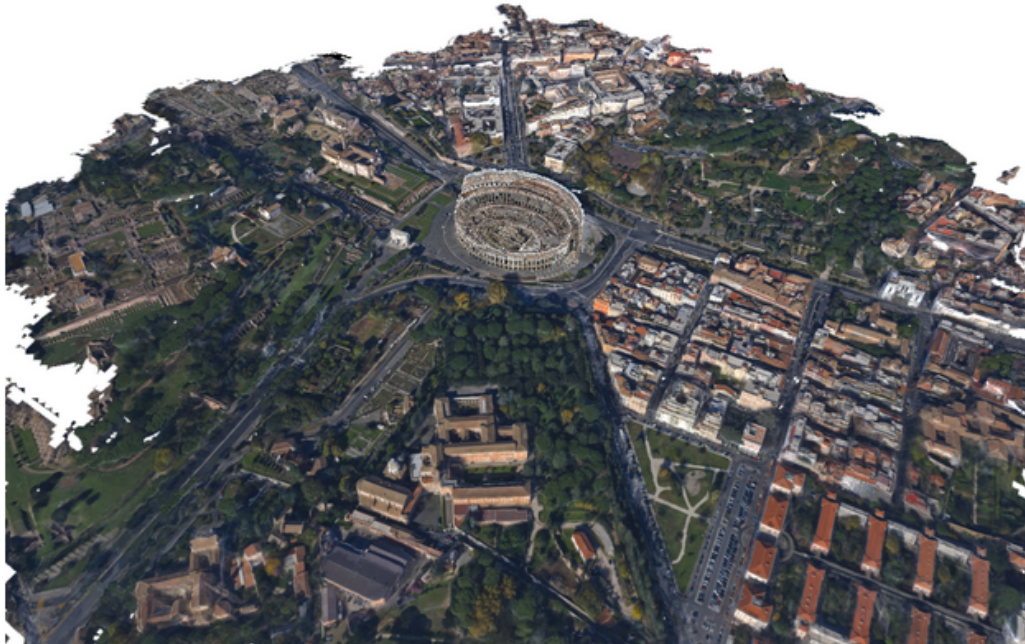


Orange Crowned Warbler

What can computer vision do today?

Reconstruction: 3D from photo collections

Colosseum, Rome, Italy



San Marco Square, Venice, Italy



Q. Shan, R. Adams, B. Curless, Y. Furukawa, and S. Seitz, [The Visual Turing Test for Scene Reconstruction](#), 3DV 2013

Reconstruction: 4D from photo collections

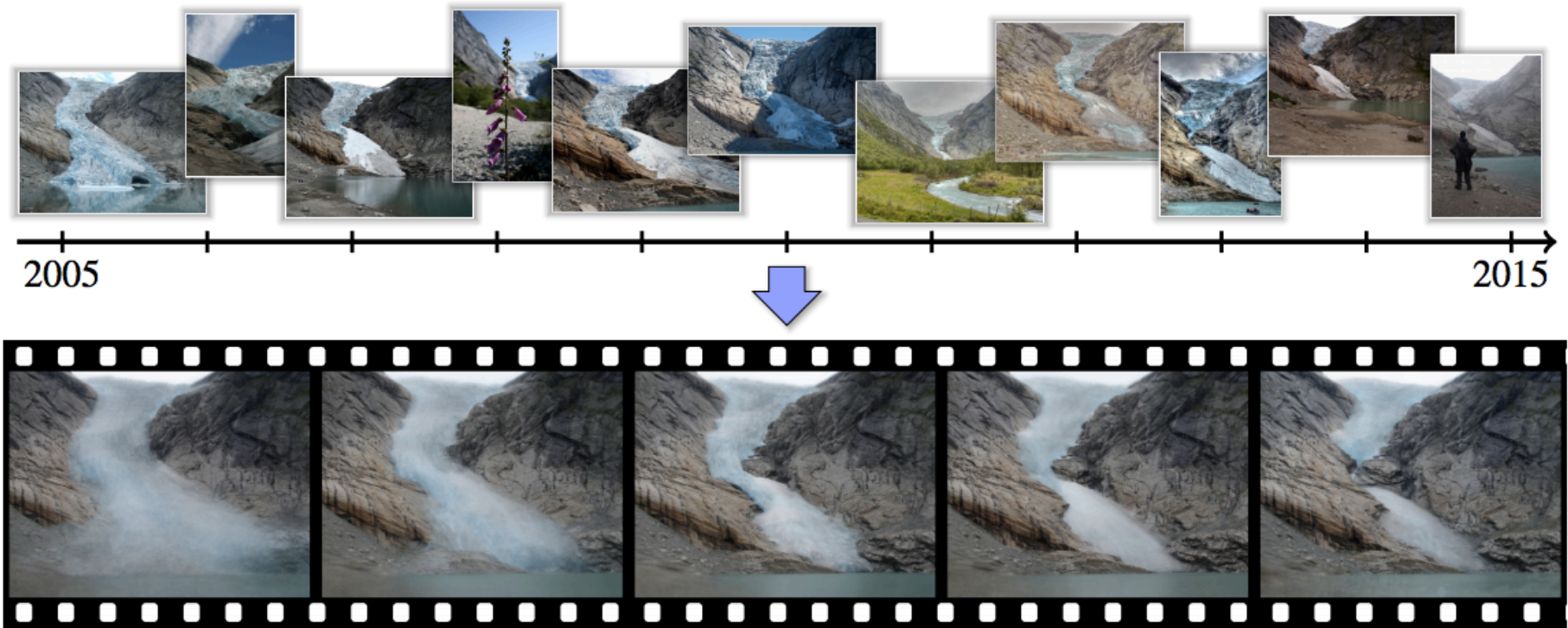


Figure 1: We mine Internet photo collections to generate time-lapse videos of locations all over the world. Our time-lapses visualize a multitude of changes, like the retreat of the Briksdalsbreen Glacier in Norway shown above. The continuous time-lapse (bottom) is computed from hundreds of Internet photos (samples on top). Photo credits: Aliento Más Allá, jirihnidek, mcxurxo, elka.cz, Juan Jesús Orío, Klaus Wißkirchen, Daikrieg, Free the image, dration and Nadav Tobias.

R. Martin-Brualla, D. Gallup, and S. Seitz, [Time-Lapse Mining from Internet Photos](#), SIGGRAPH 2015

[YouTube Video](#)

Reconstruction: 4D from depth cameras



Figure 1: Real-time reconstructions of a moving scene with DynamicFusion; both the person and the camera are moving. The initially noisy and incomplete model is progressively denoised and completed over time (left to right).

R. Newcombe, D. Fox, and S. Seitz, [DynamicFusion: Reconstruction and Tracking of Non-rigid Scenes in Real-Time](#), CVPR 2015

[YouTube Video](#)

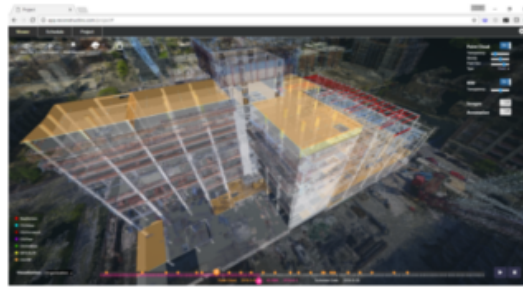
Reconstruction in construction industry

RECONSTRUCT INTEGRATES REALITY AND PLAN



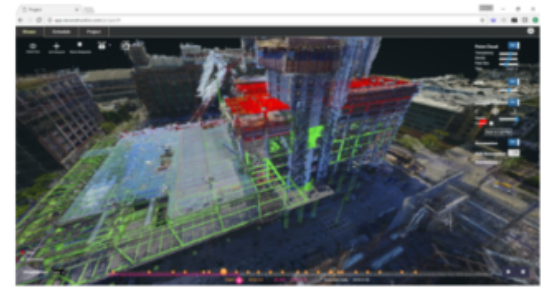
Visual Asset Management

Reconstruct 4D point clouds and organize images and videos from smartphones, time-lapse cameras, and drones around the project schedule. View, annotate, and share anywhere with a web interface.



4D Visual Production Models

Integrate 4D point clouds with 4D BIM, review "who does what work at what location" on a daily basis and improve coordination and communication among project teams.

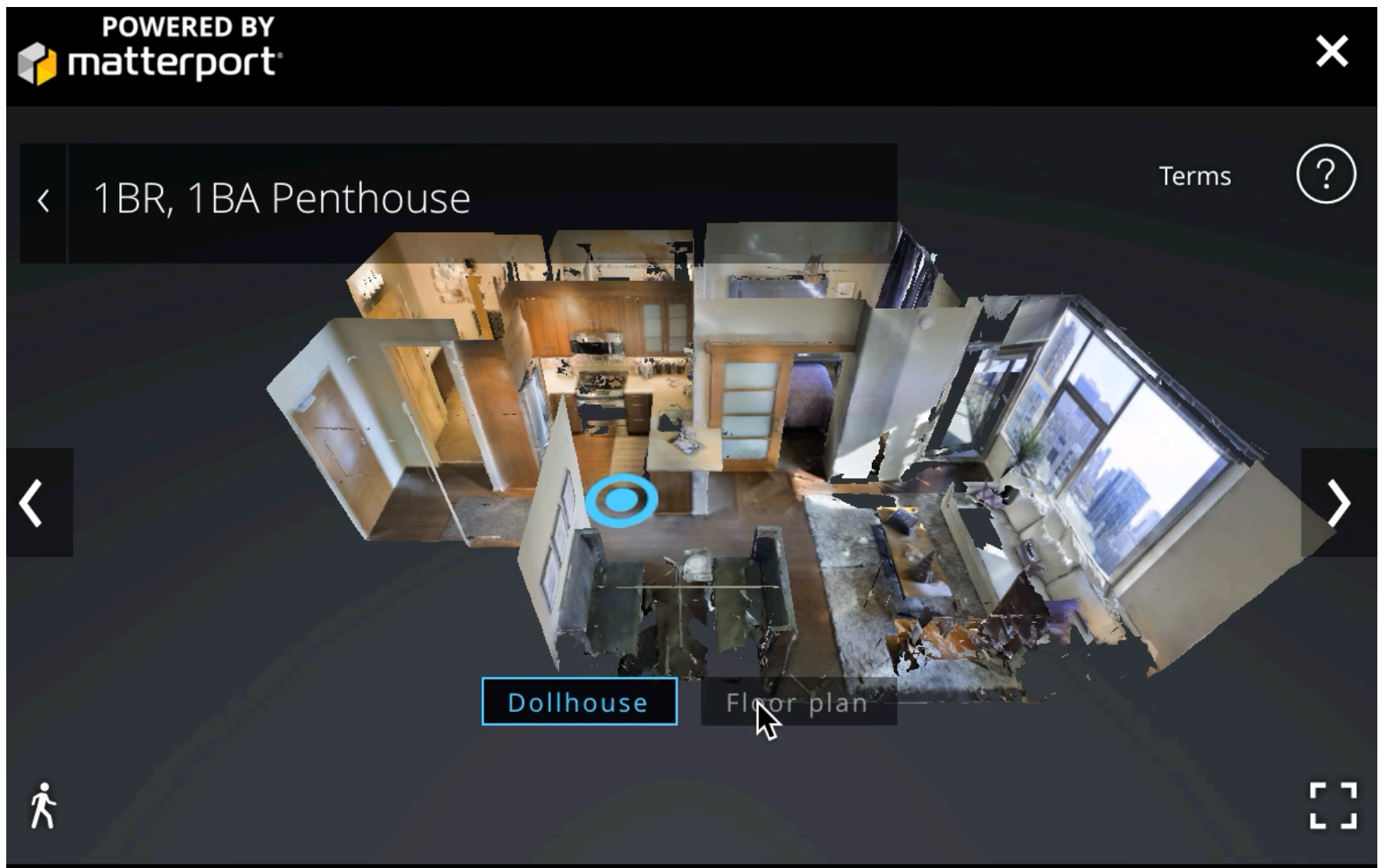


Predictive Visual Data Analytics

Analyze actual progress deviations by comparing Reality and Plan and predict risk with respect to the execution of the look-ahead schedule for each project location, to offer your project team with an opportunity to tap off potential delays before they surface on your jobsite.

reconstructinc.com

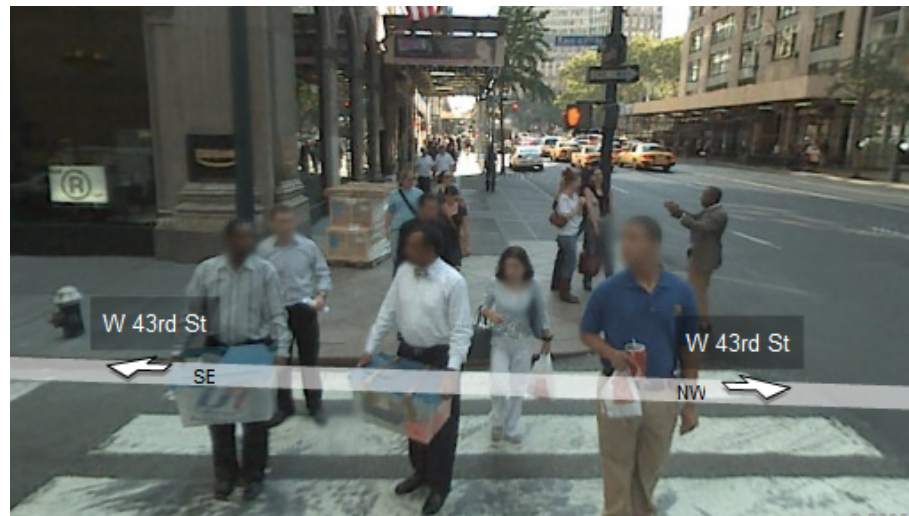
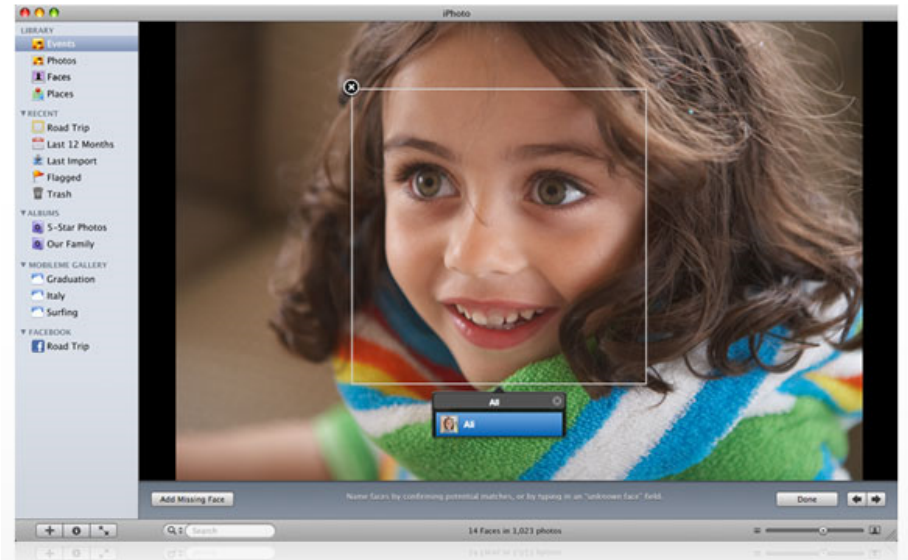
Applications



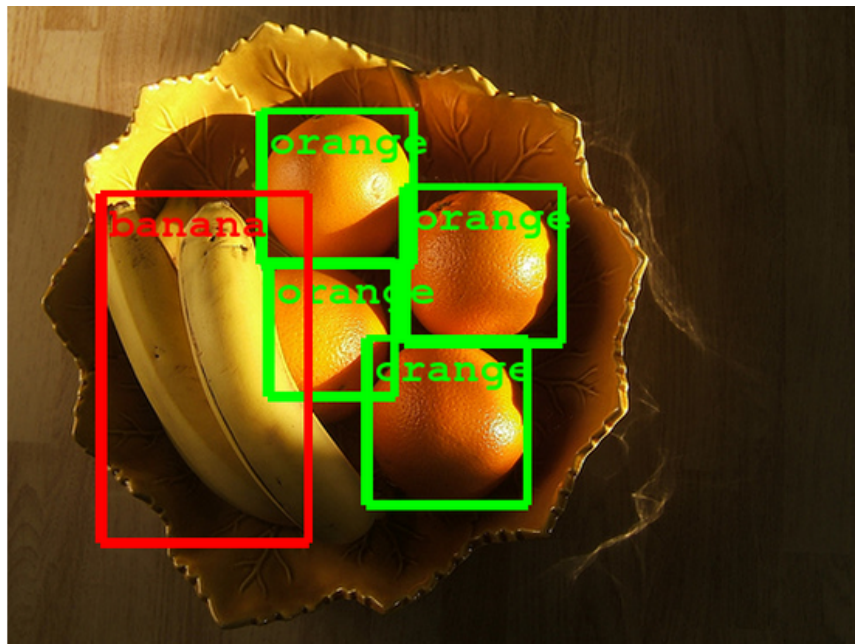
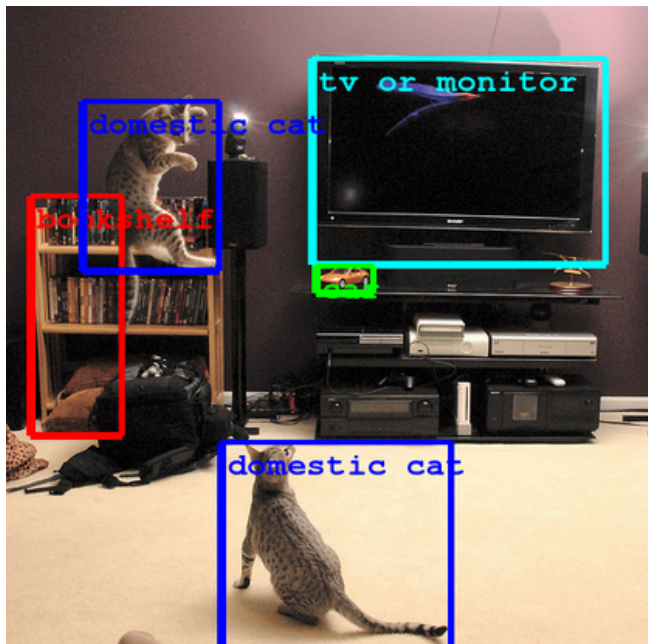
Recognition: "Simple" patterns



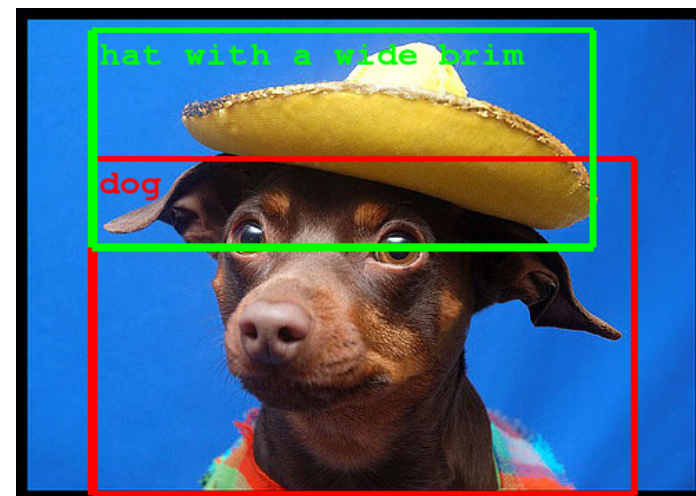
Recognition: Faces



Recognition: General categories

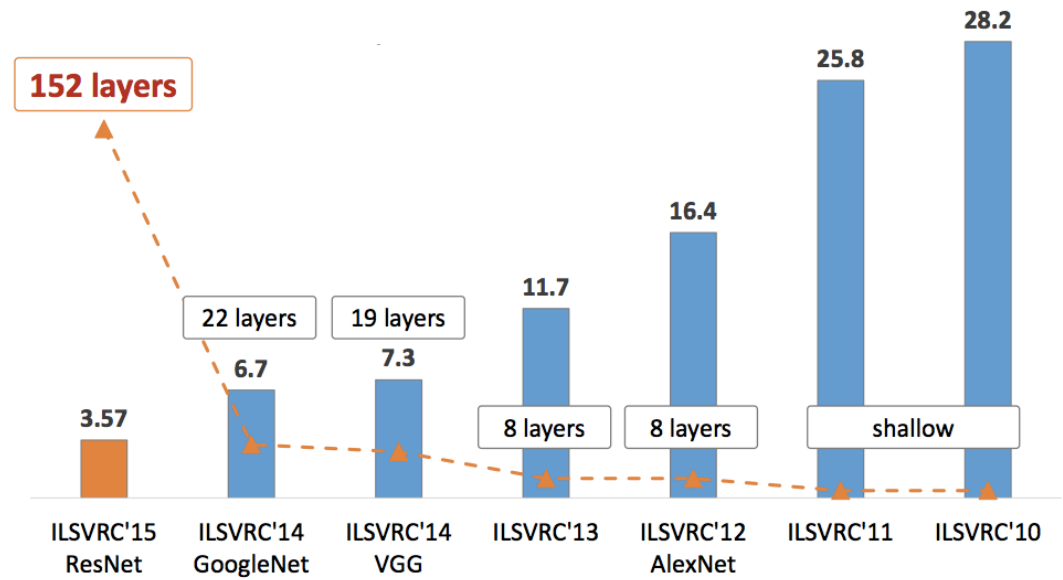
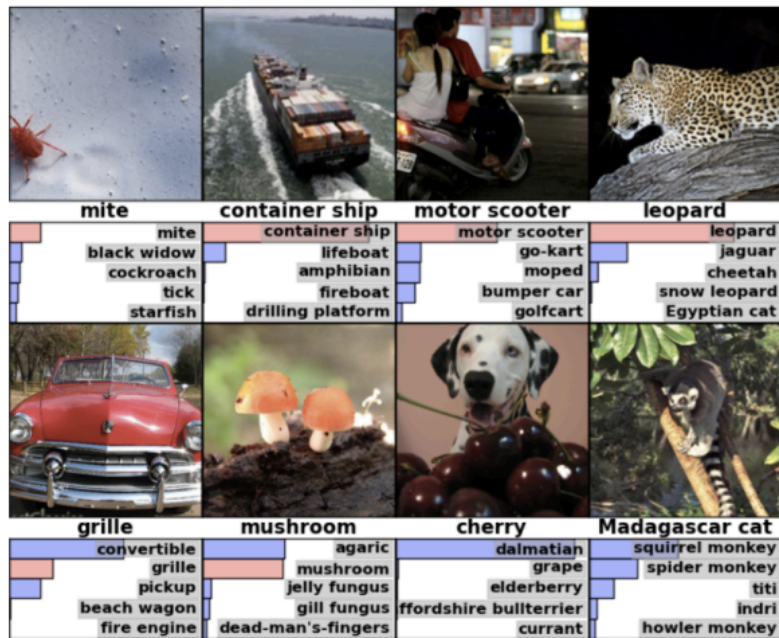


- [Computer Eyesight Gets a Lot More Accurate](#), NY Times Bits blog, August 18, 2014
- [Building A Deeper Understanding of Images](#), Google Research Blog, September 5, 2014

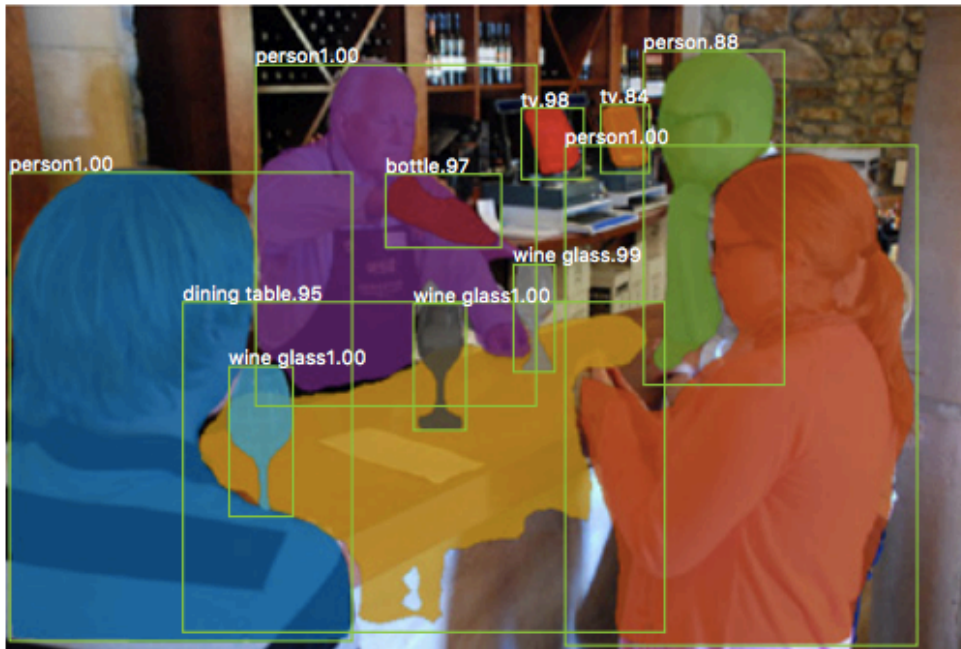


Recognition: General categories

- [ImageNet challenge](#)



Object detection, instance segmentation



K. He, G. Gkioxari, P. Dollar, and R. Girshick, [Mask R-CNN](#),
ICCV 2017 (Best Paper Award)

Source: L. Lazebnik

Image generation

- Faces: 1024x1024 resolution, CelebA-HQ dataset

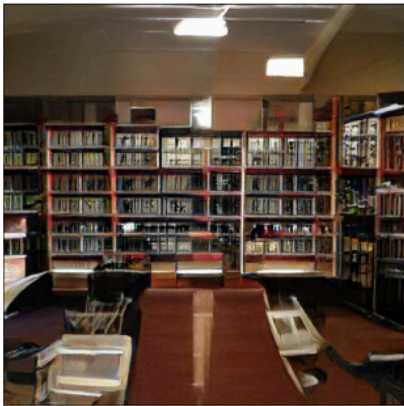
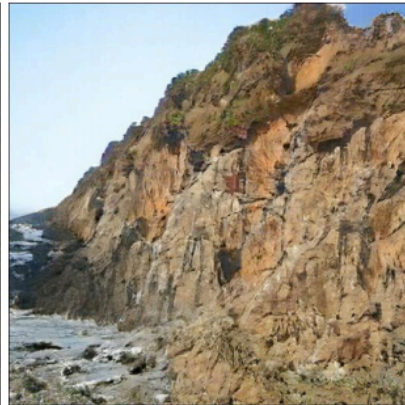


T. Karras, T. Aila, S. Laine, and J. Lehtinen, [Progressive Growing of GANs for Improved Quality, Stability, and Variation](#), ICLR 2018

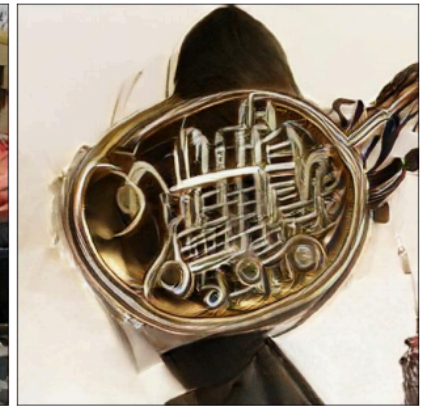
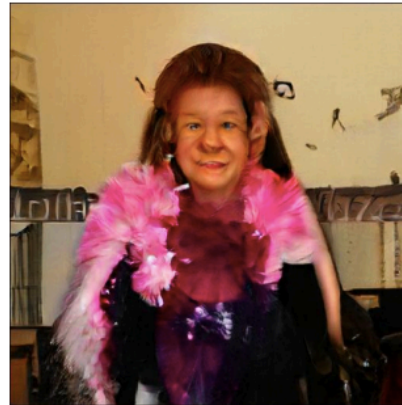
Image generation

- BigGAN: 512 x 512 resolution, ImageNet

Easy classes



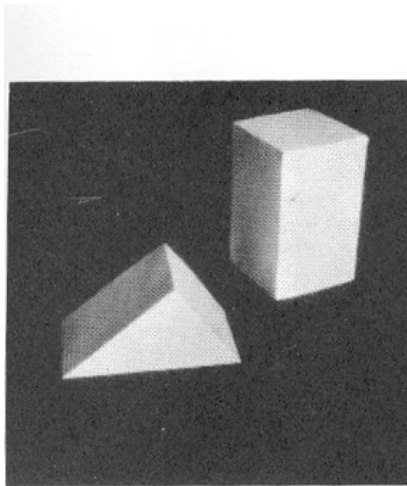
Difficult classes



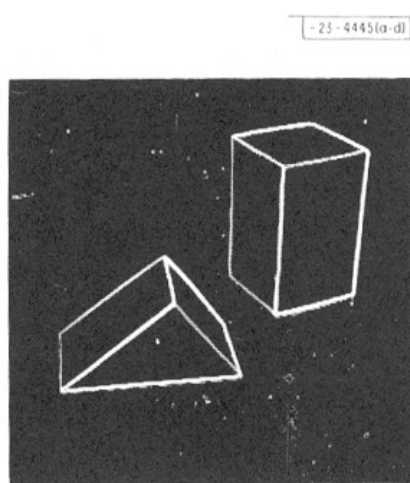
A. Brock, J. Donahue, K. Simonyan, [Large scale GAN training for high fidelity natural image synthesis](#), arXiv 2018

Source: L. Lazebnik

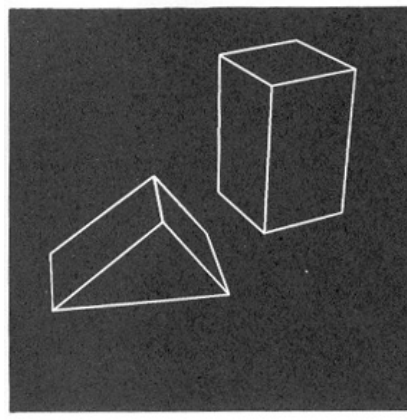
Origins of computer vision



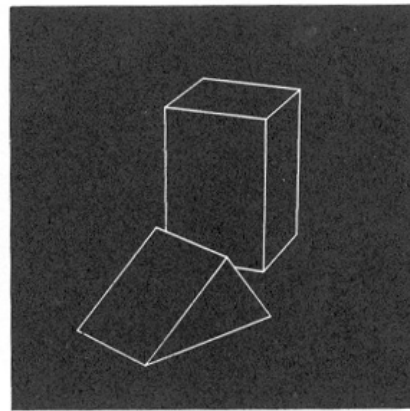
(a) Original picture.



(b) Differentiated picture.



(c) Line drawing.



(d) Rotated view.

[L. G. Roberts *Machine Perception of Three Dimensional Solids*](#)

Origins of computer vision

MASSACHUSETTS INSTITUTE OF TECHNOLOGY
PROJECT MAC

Artificial Intelligence Group
Vision Memo. No. 100.

July 7, 1966

THE SUMMER VISION PROJECT

Seymour Papert

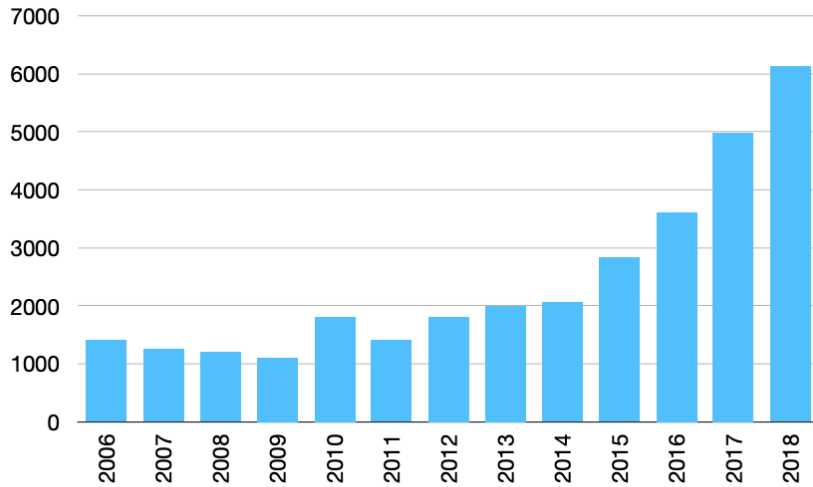
The summer vision project is an attempt to use our summer workers effectively in the construction of a significant part of a visual system. The particular task was chosen partly because it can be segmented into sub-problems which will allow individuals to work independently and yet participate in the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

Six decades of computer vision

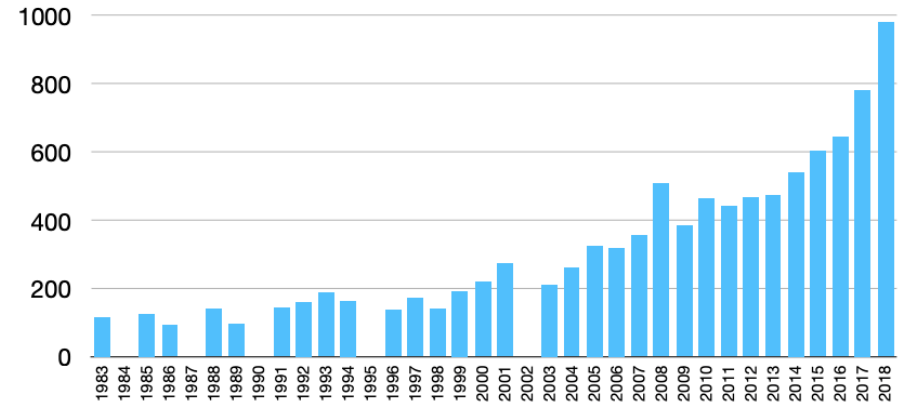
- 1960s: Beginnings in artificial intelligence, image processing and pattern recognition
- 1970s: Foundational work on image formation: Horn, Koenderink, Longuet-Higgins ...
- 1980s: Vision as applied mathematics: geometry, multi-scale analysis, probabilistic modeling, control theory, optimization
- 1990s: Geometric analysis largely completed, vision meets graphics, statistical learning approaches resurface
- 2000s: Significant advances in visual recognition
- 2010s: Progress continues, aided by the availability of large amounts of visual data and massive computing power. Deep learning has become pre-eminent

Growth of the field

CVPR Attendance



CVPR Papers



[Source](#)

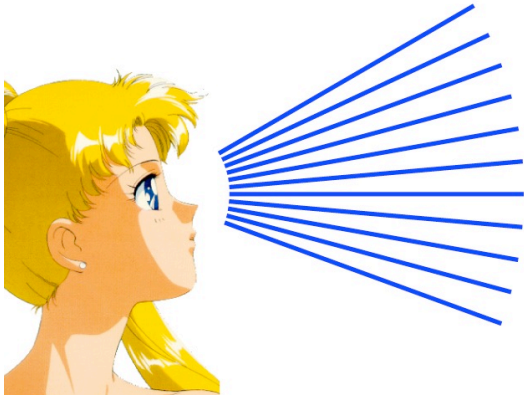
[Long list of corporate sponsors](#)

Course overview

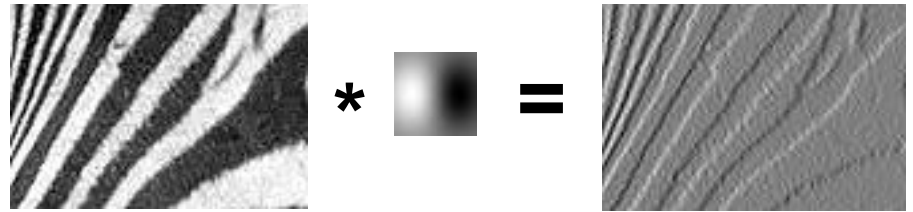
- I. Early vision: Image formation and processing
- II. Mid-level vision: Grouping and fitting
- III. Multi-view geometry
- IV. Recognition
- V. Additional topics

I. Early vision

Basic image formation and processing



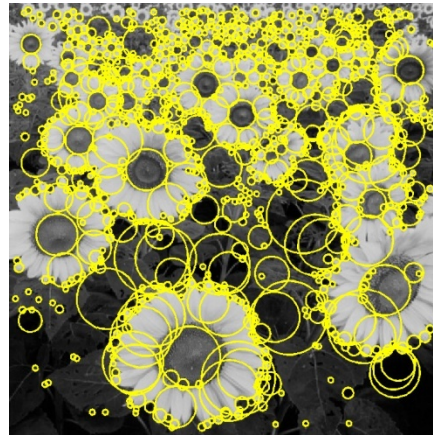
Cameras and sensors
Light and color



Linear filtering
Edge detection



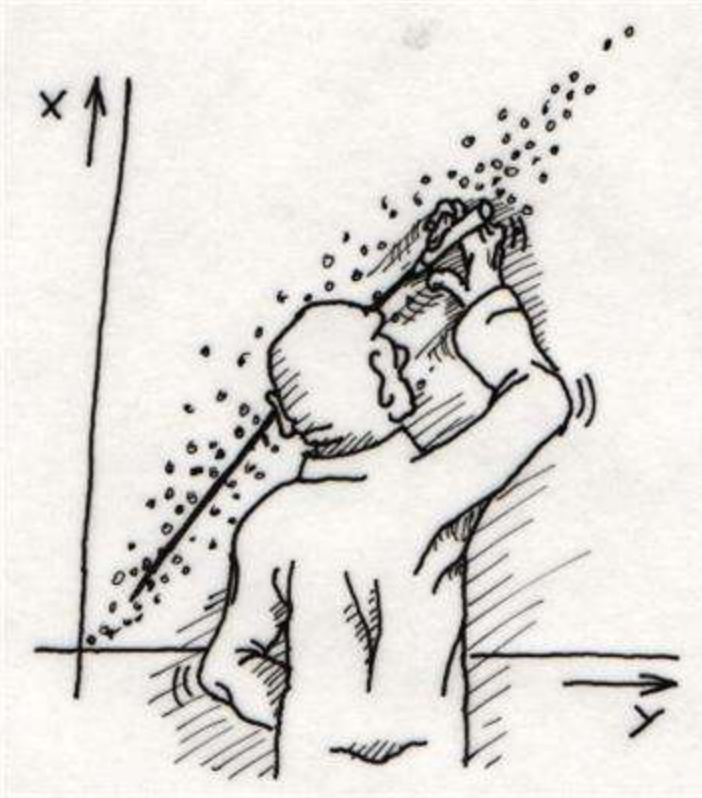
Feature extraction



Optical flow

II. “Mid-level vision”

Fitting and grouping

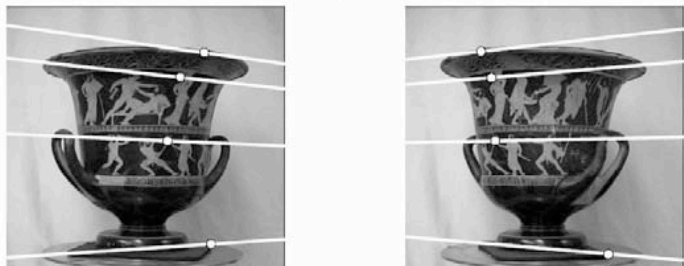


Fitting: Least squares
Voting methods



Alignment

III. Multi-view geometry



Epipolar geometry

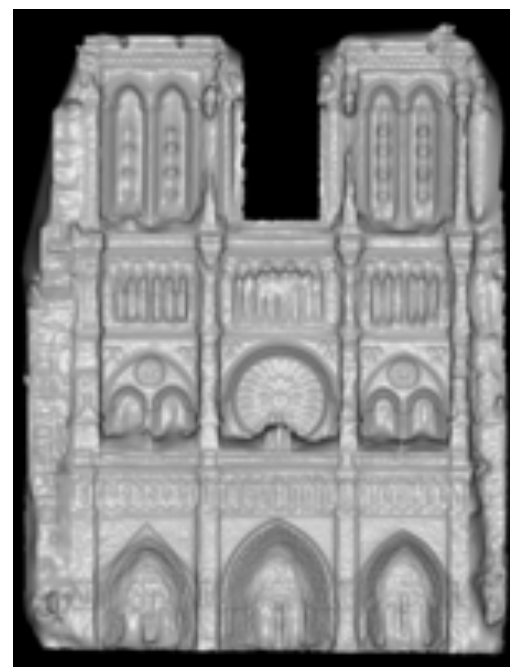


Two-view stereo



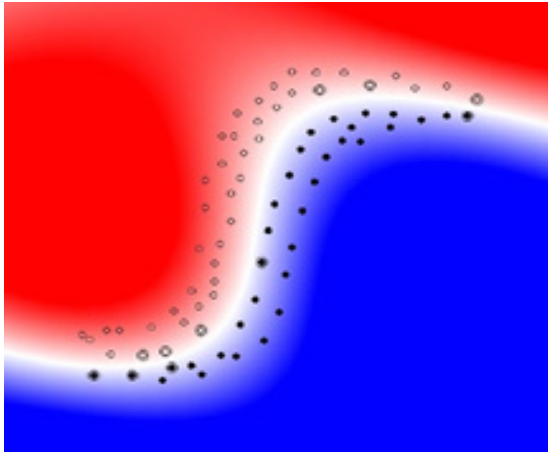
Драконъ, видимый подъ различными углами зрѣнія
По гравюру на мѣди изъ „Oculus artificialis teleiopicus“ Цана. 1702 года.

Structure from motion

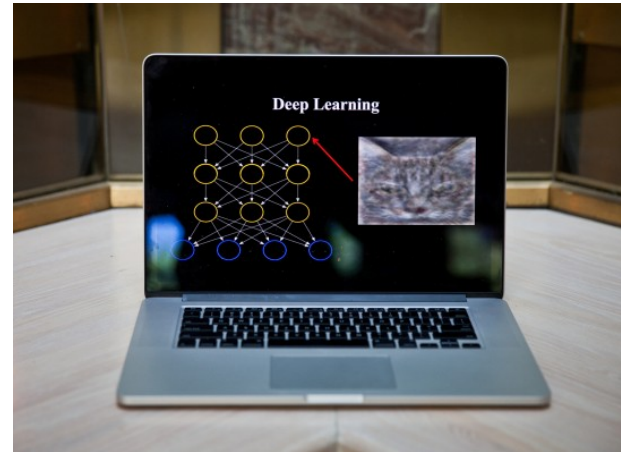


Multi-view stereo

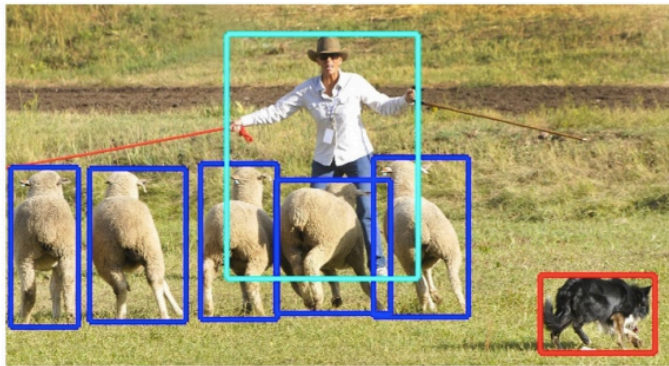
IV. Recognition



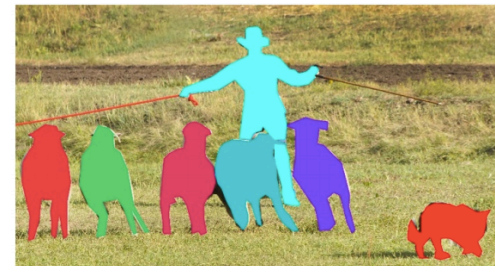
Basic classification



Deep learning



Object detection



Segmentation

