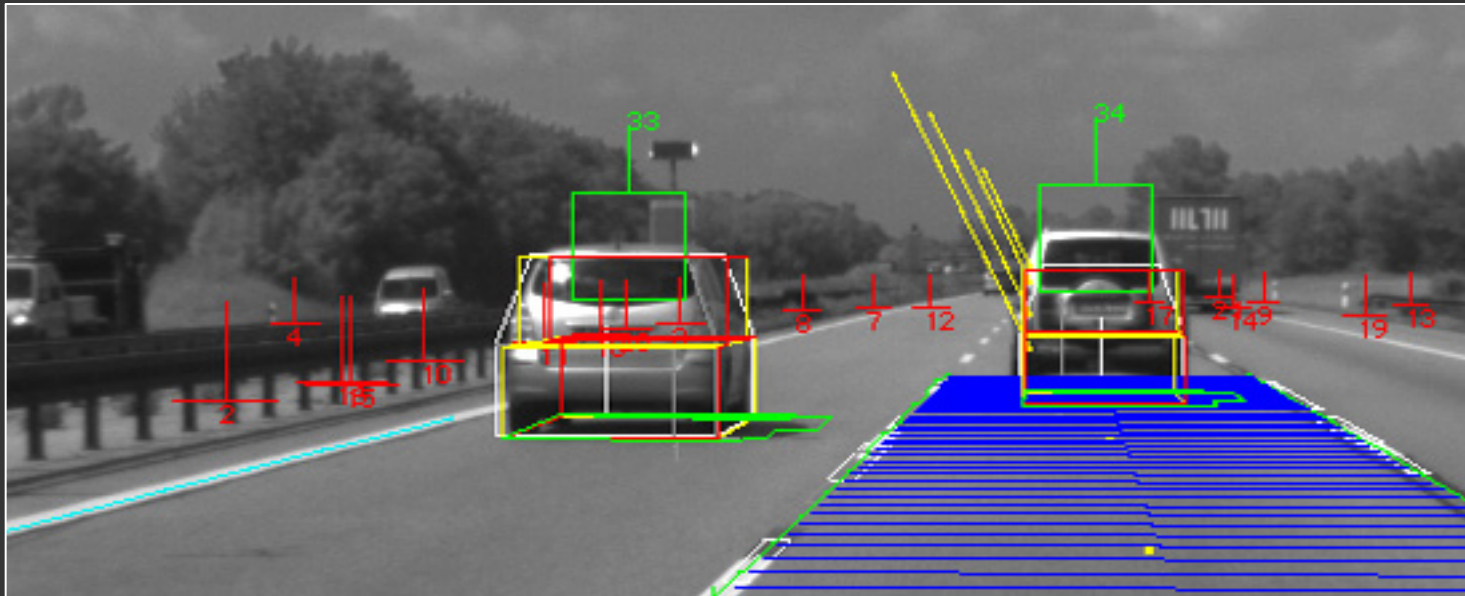


Special Topics in Computer Vision

2P 2019



Course webpage URL:

opilab.utb.edu.co/topics-computer-vision/

Contact Details

Instructor: Andrés G. Marrugo, PhD

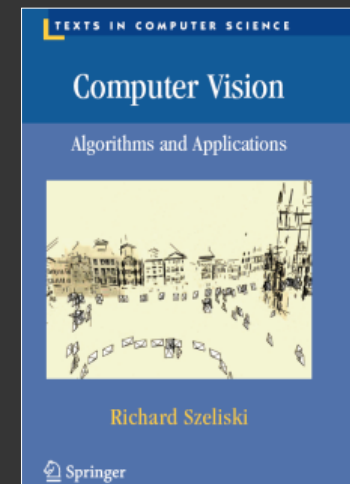
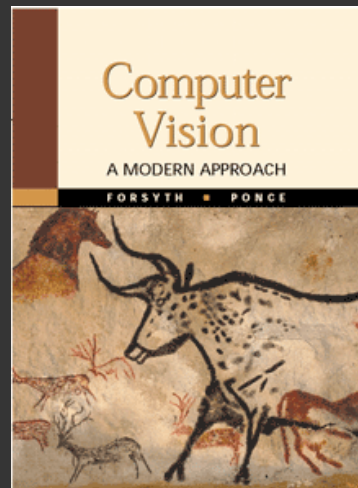
Email: agmarrugo@utb.edu.co Twitter: [@agmarrugo](https://twitter.com/agmarrugo)

Course website: <http://opilab.utb.edu.co/topics-computer-vision/>

Suggested textbooks (for fundamentals):

Forsyth & Ponce, *Computer Vision: A Modern Approach*

Richard Szeliski, [Computer Vision: Algorithms and Applications](#)
(available online)



About the course

We will select several topics in Computer vision, read and analyze related papers, and produce a simple implementation.

- Reading assignments (~6, 40%).
- Oral presentation (30%)
- Final project with written paper* (30%)

*For some, the paper is to be submitted to a conference.

Please refer to the syllabus.

Disclosure

Parts of this course have been prepared based on the publicly available information from the courses:

- Special Topics in Computer Vision by Noah Snavely at Cornell University.
- Computer Vision by Svetlana Lazebnik at University of Illinois at Urbana-Champaign.

Credit is given to the corresponding author.

The goal of computer vision

- To extract “meaning” from pixels



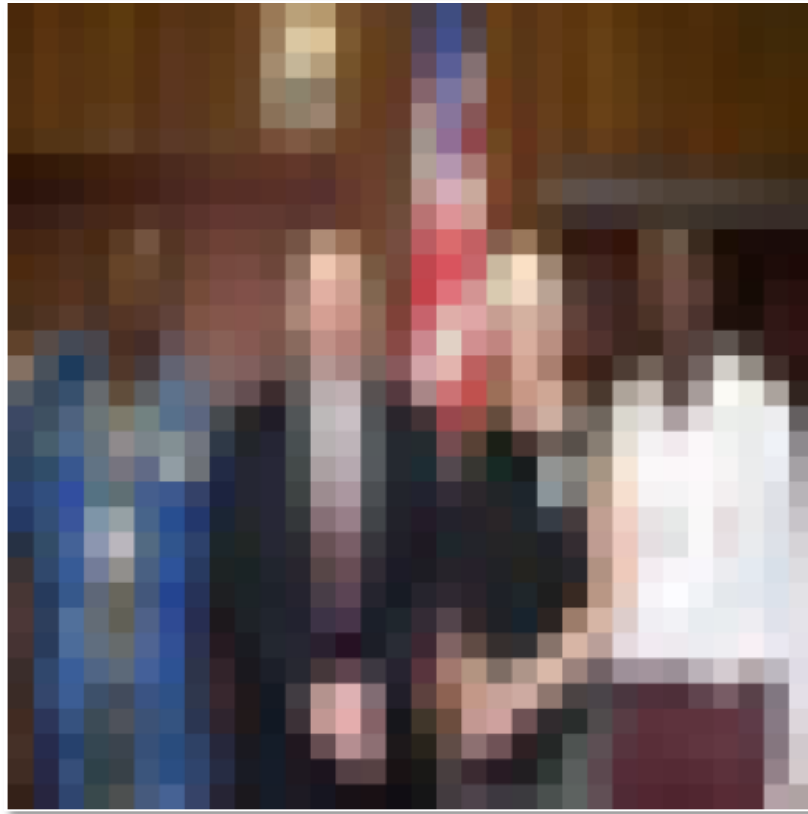
What we see

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

The goal of computer vision

- To extract “meaning” from pixels

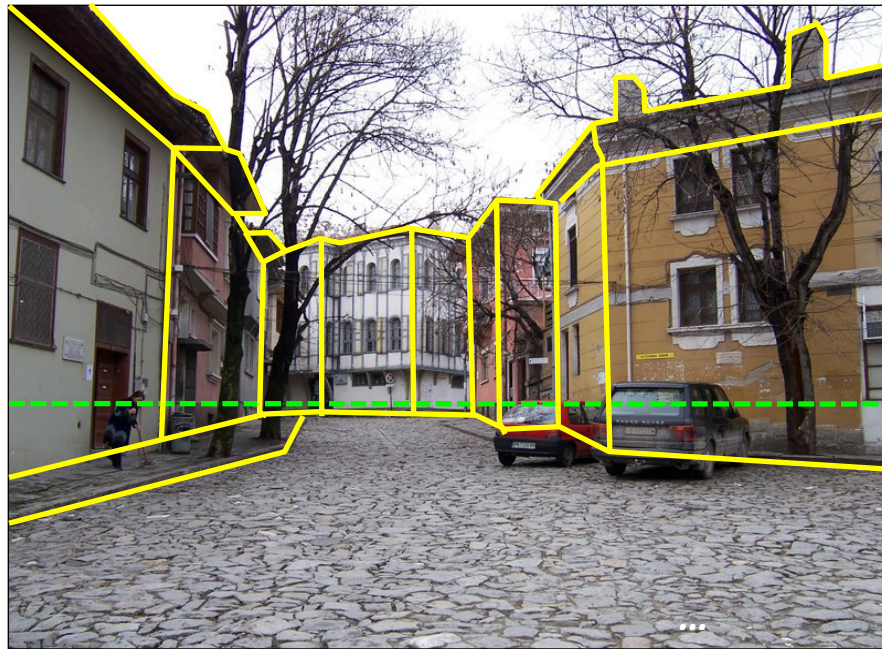


Humans are remarkably good at this...

What kind of information can be extracted from an image?

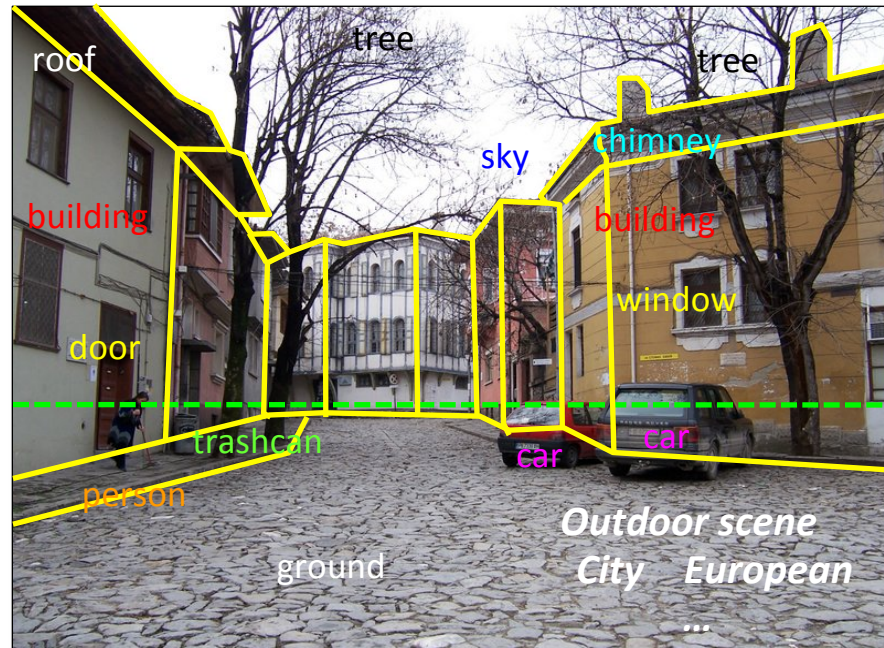


What kind of information can be extracted from an image?



Geometric information

What kind of information can be extracted from an image?

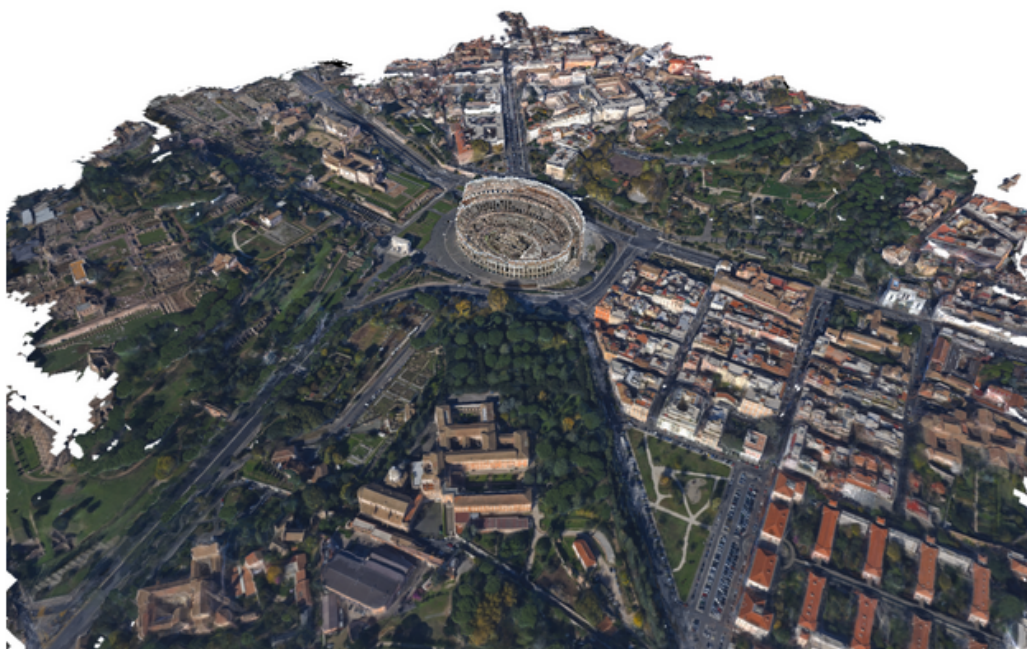


Geometric information
Semantic information

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

[YouTube Video](#)

Reconstruction: 4D from photo collections

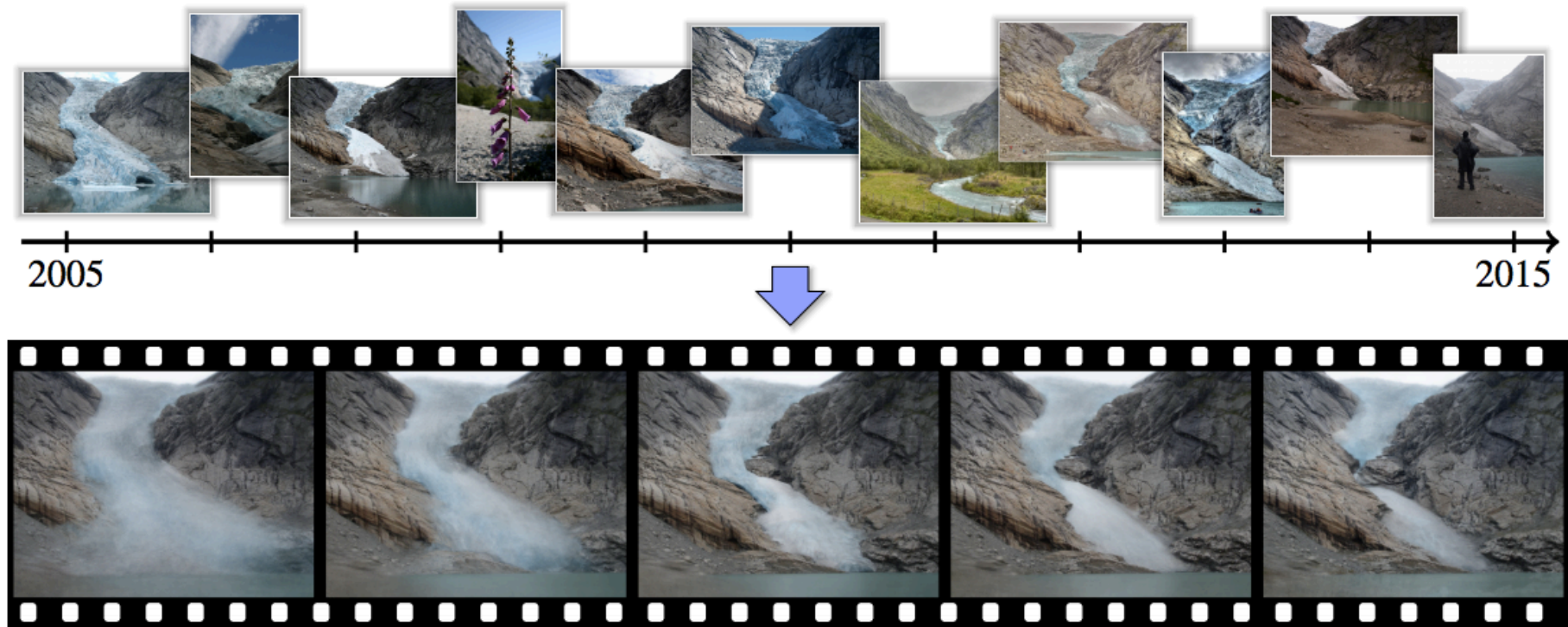


Figure 1: We mine Internet photo collections to generate time-lapse videos of locations all over the world. Our time-lapses visualize a multitude of changes, like the retreat of the Briksdalsbreen Glacier in Norway shown above. The continuous time-lapse (bottom) is computed from hundreds of Internet 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



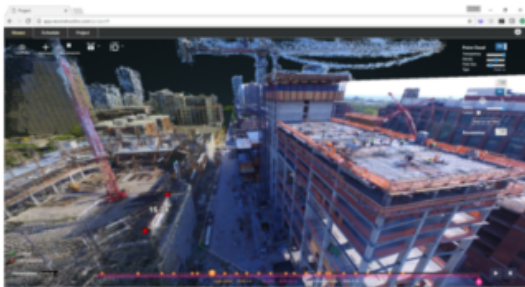
Figure 1: Real-time reconstructions of a moving scene with DynamicFusion; both the person and the camera are moving. The initially noisy and incomplete model is progressively denoised and completed over time (left to right).

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

[YouTube Video](#)

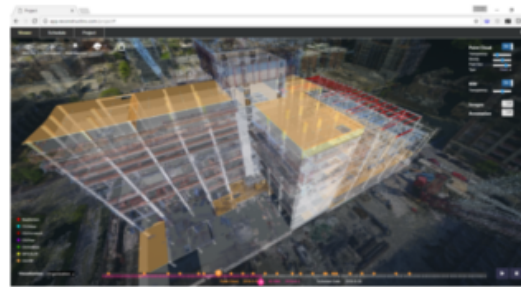
Reconstruction in construction industry

RECONSTRUCT INTEGRATES REALITY AND PLAN



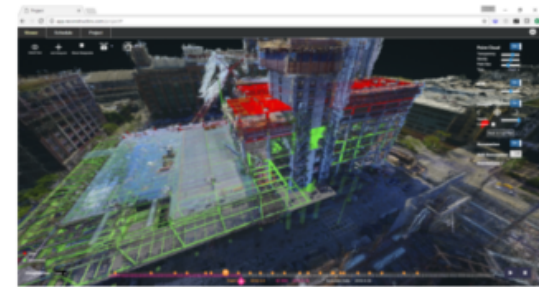
Visual Asset Management

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.

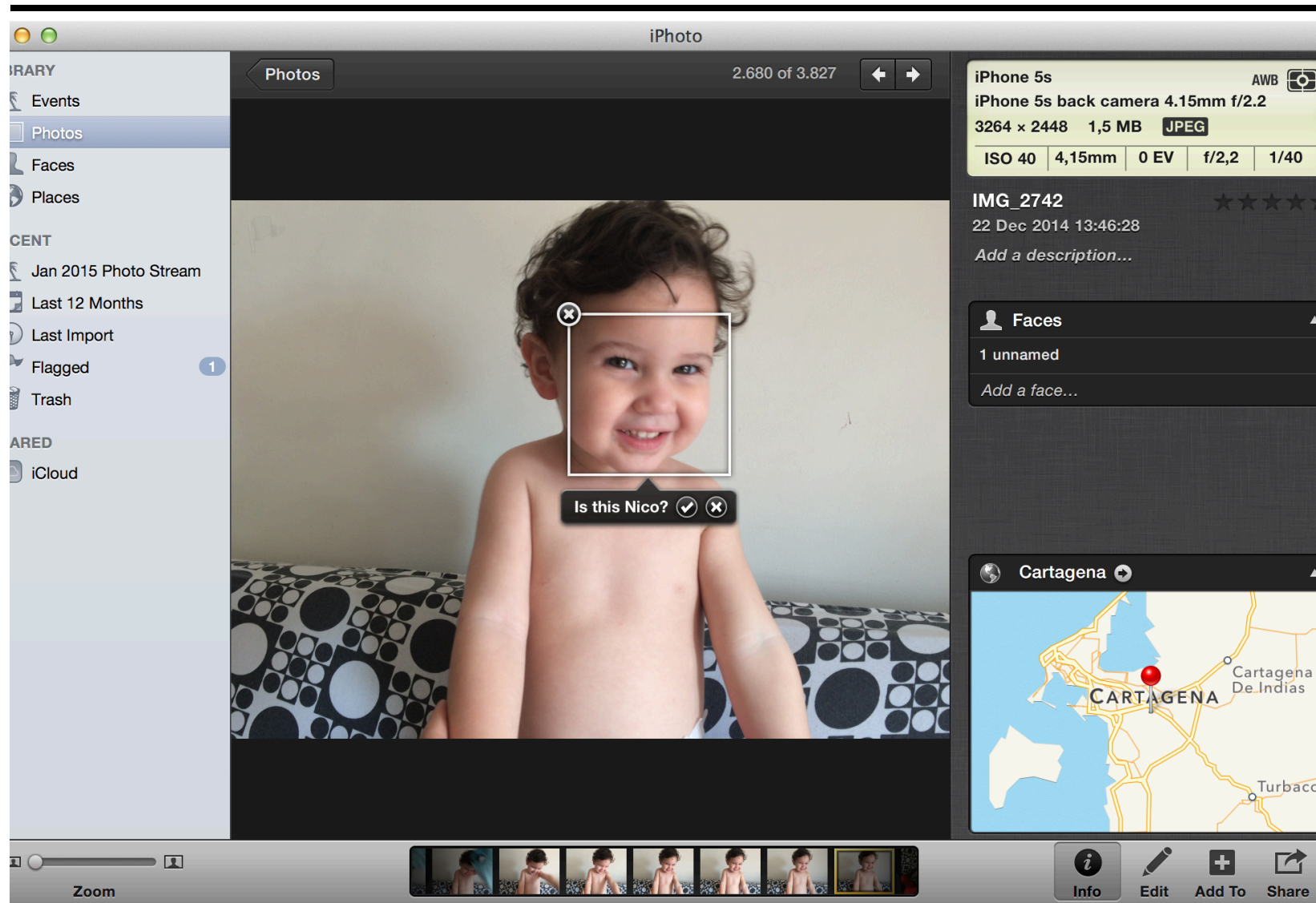
reconstructinc.com

Source: D. Hoiem

Recognition: "Simple" patterns



Face recognition: Apple iPhoto software



Recognition: Apple Face ID



Concerns about face recognition



[Beijing bets on facial recognition in a big drive for total surveillance](#) – Washington Post, 1/8/2018

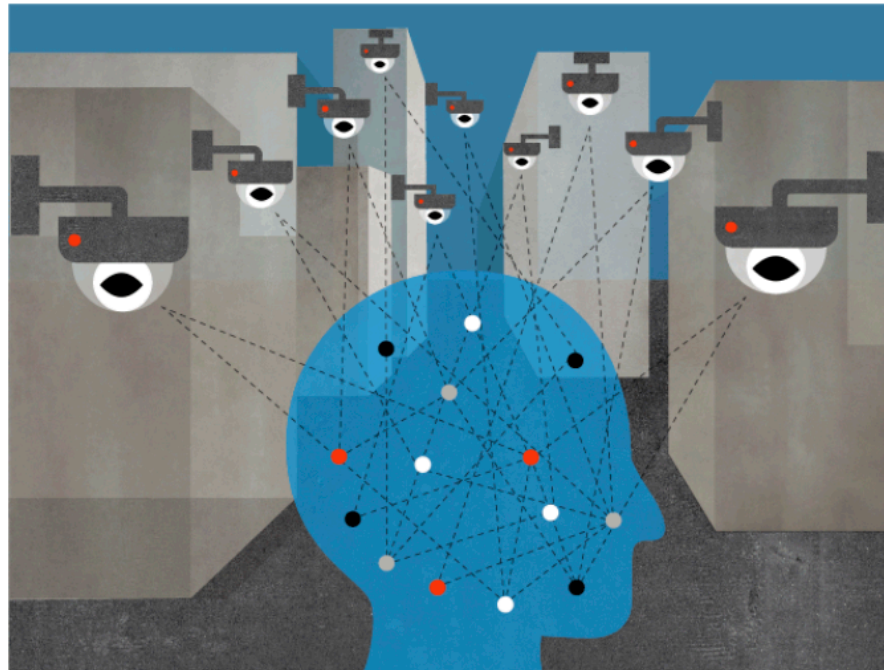
Concerns about face recognition

ANNALS OF TECHNOLOGY DECEMBER 17, 2018 ISSUE

THE
NEW YORKER

SHOULD WE BE WORRIED ABOUT COMPUTERIZED FACIAL RECOGNITION?

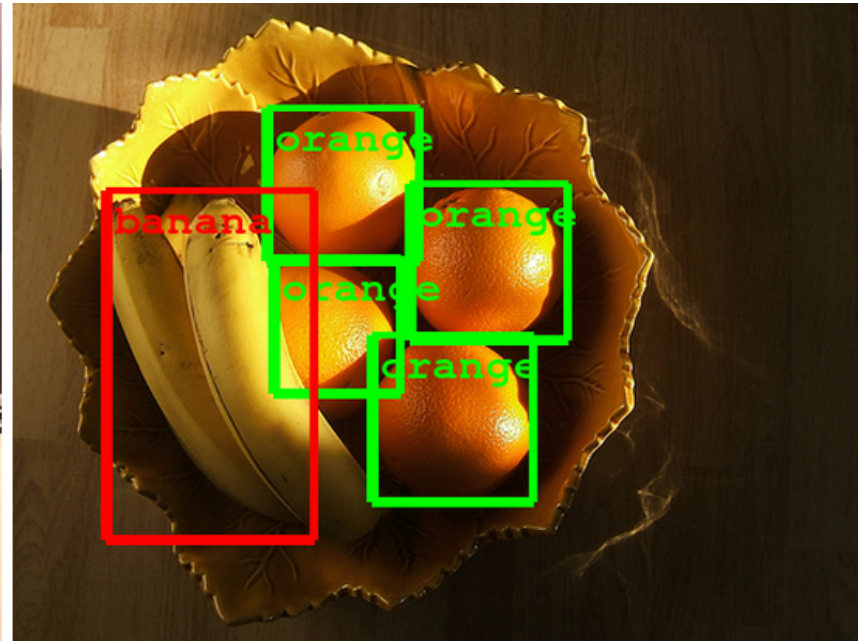
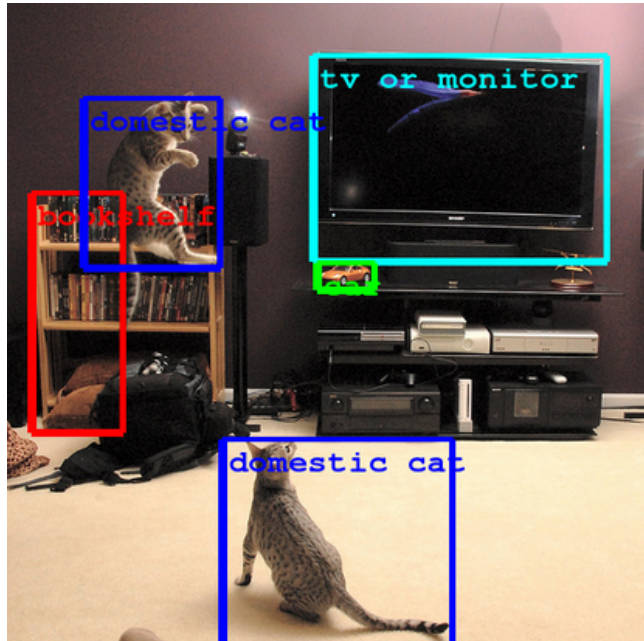
The technology could revolutionize policing, medicine, even agriculture—but its applications can easily be weaponized.



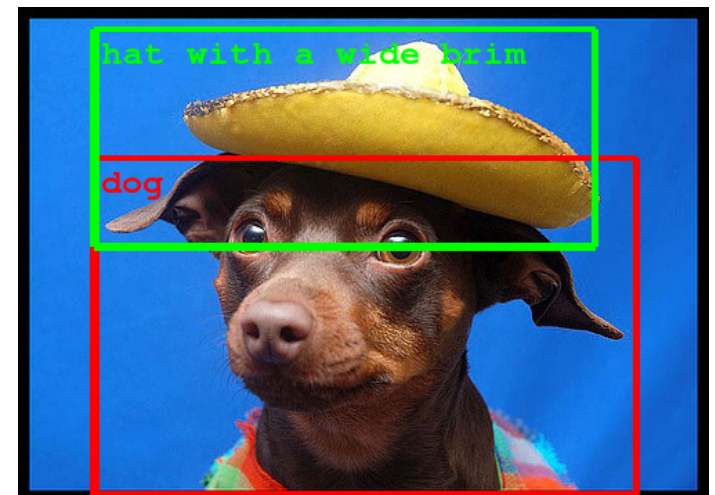
Many U.S. cities won't disclose their police departments' surveillance methods.

<https://www.newyorker.com/magazine/2018/12/17/should-we-be-worried-about-computerized-facial-recognition>

Recognition: General categories

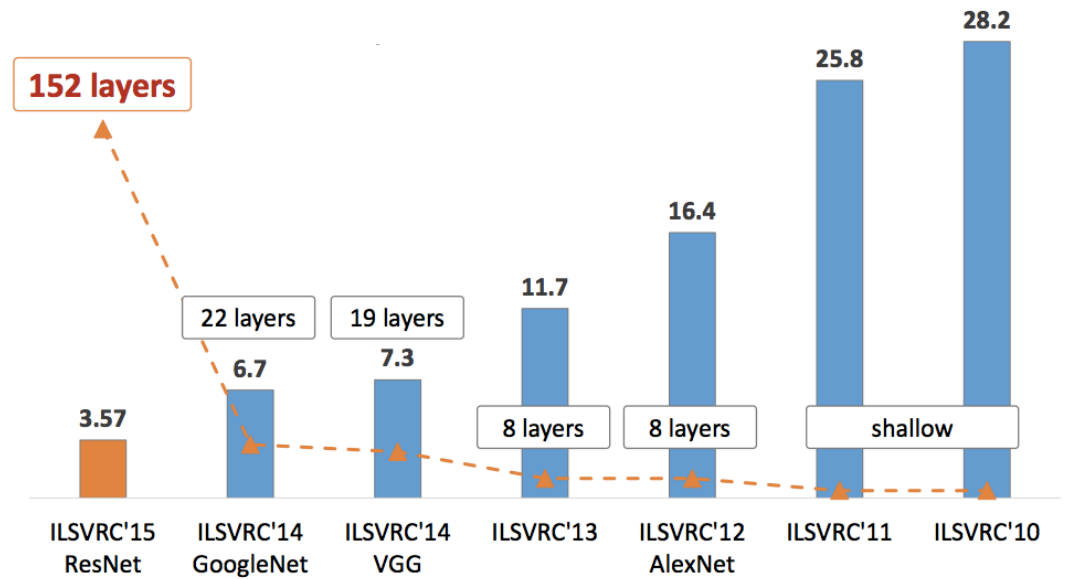
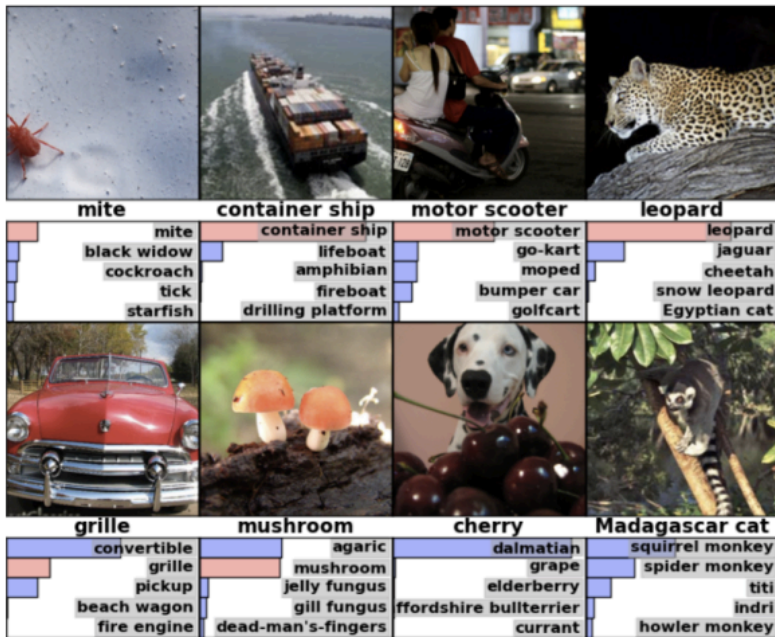


- [Computer Eyesight Gets a Lot More Accurate](#), NY Times Bits blog, August 18, 2014
- [Building A Deeper Understanding of Images](#), Google Research Blog, September 5, 2014

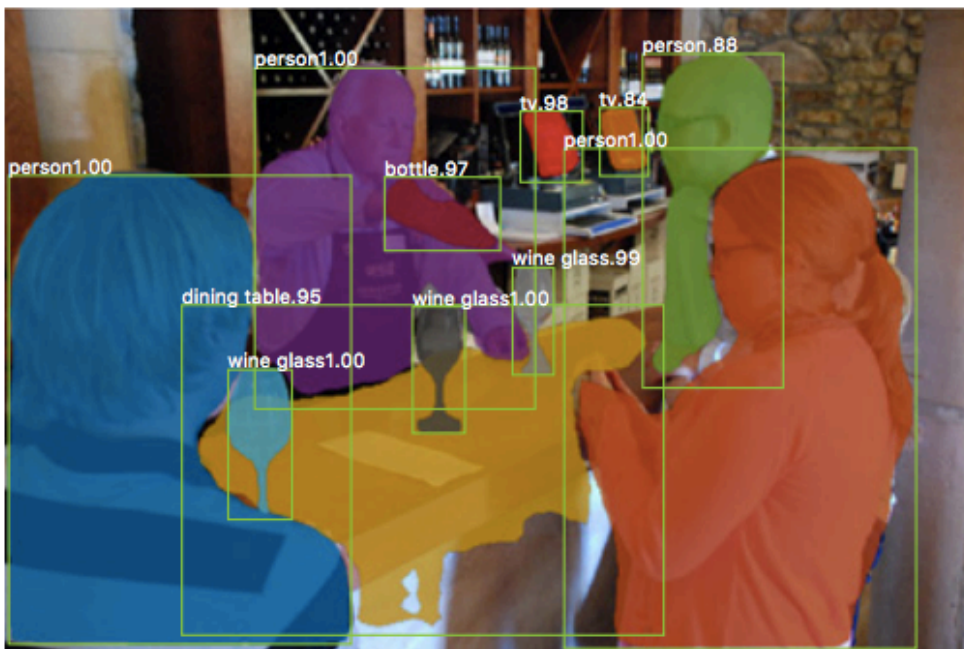


Recognition: General categories

- [ImageNet challenge](#)



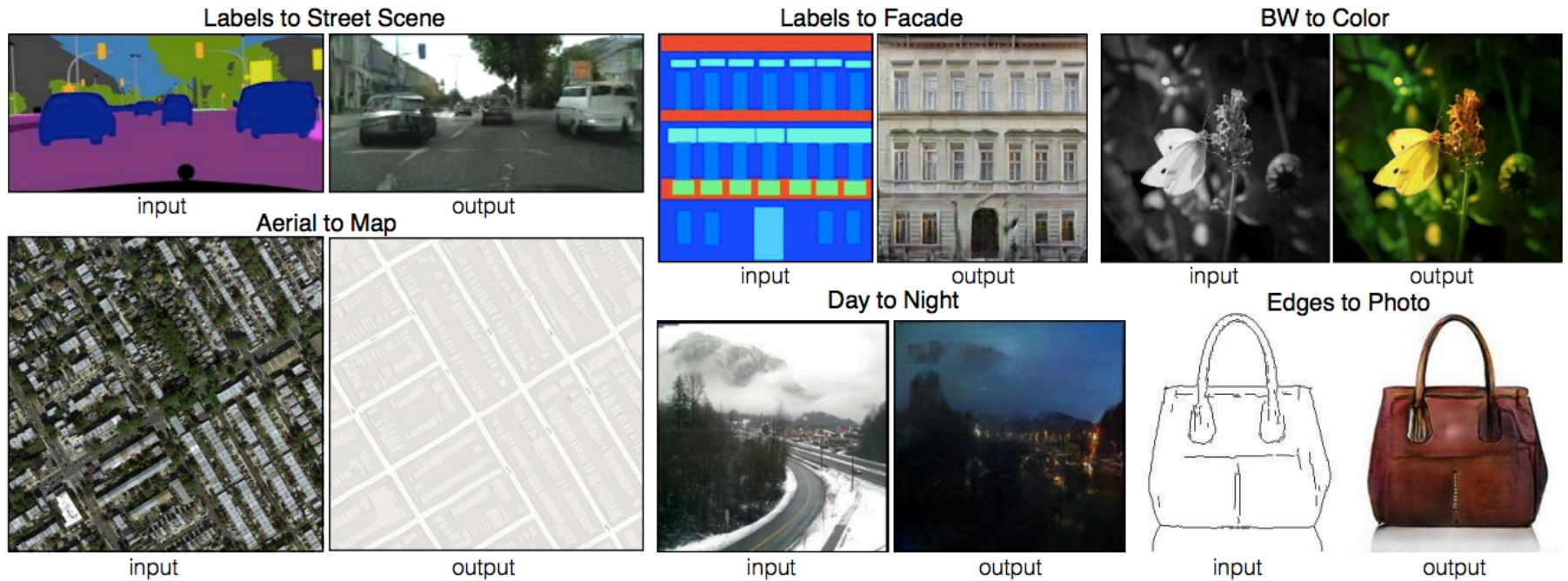
Object detection, instance segmentation



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

Image generation

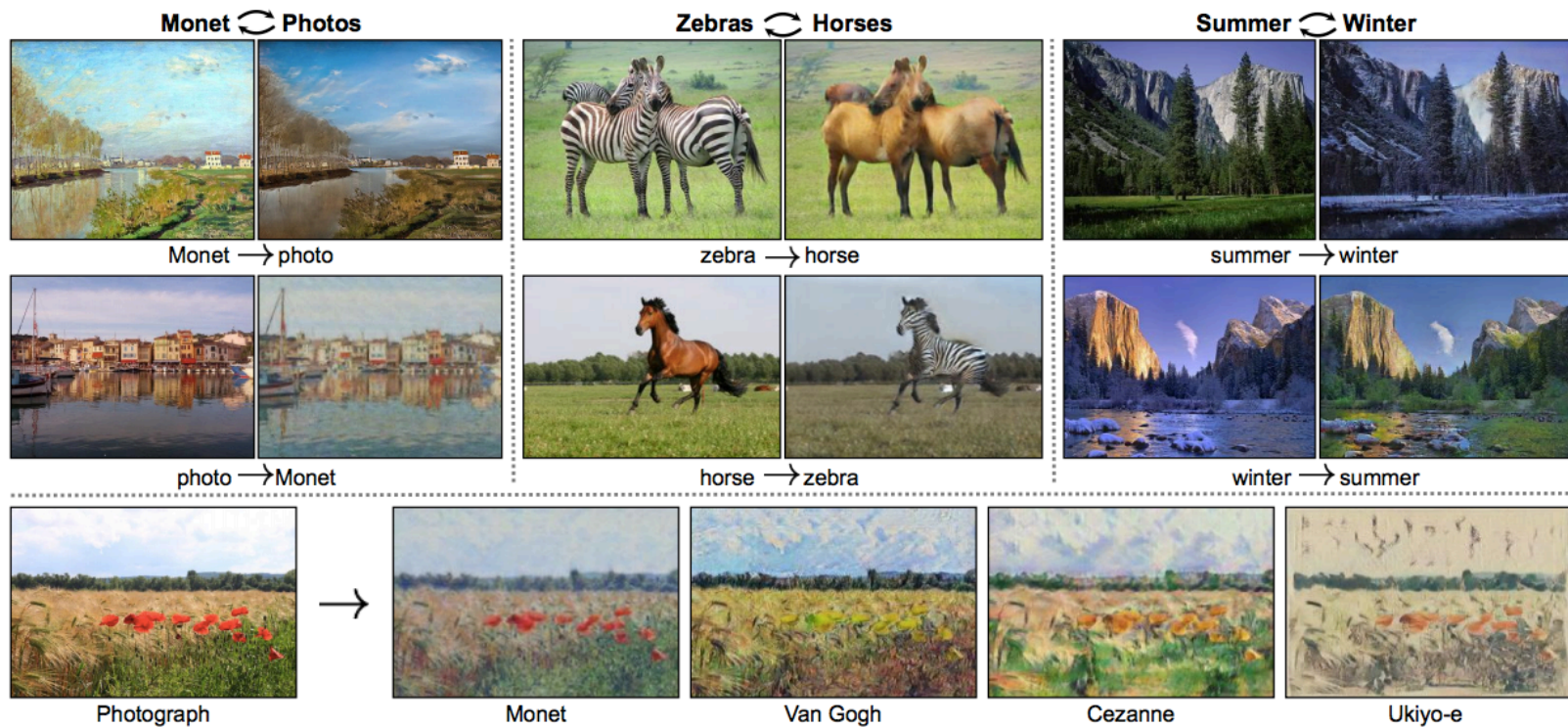
- Image-to-image translation



P. Isola, J.-Y. Zhu, T. Zhou, A. Efros,
[Image-to-Image Translation with Conditional Adversarial Networks](#), CVPR 2017

Image generation

- Unpaired image-to-image translation



J.-Y. Zhu, T. Park, P. Isola, A. Efros, [Unpaired Image-to-Image Translation Using Cycle-Consistent Adversarial Networks](#), ICCV 2017

Unsupervised image-to-image translation



Figure 4: Dog breed translation results.

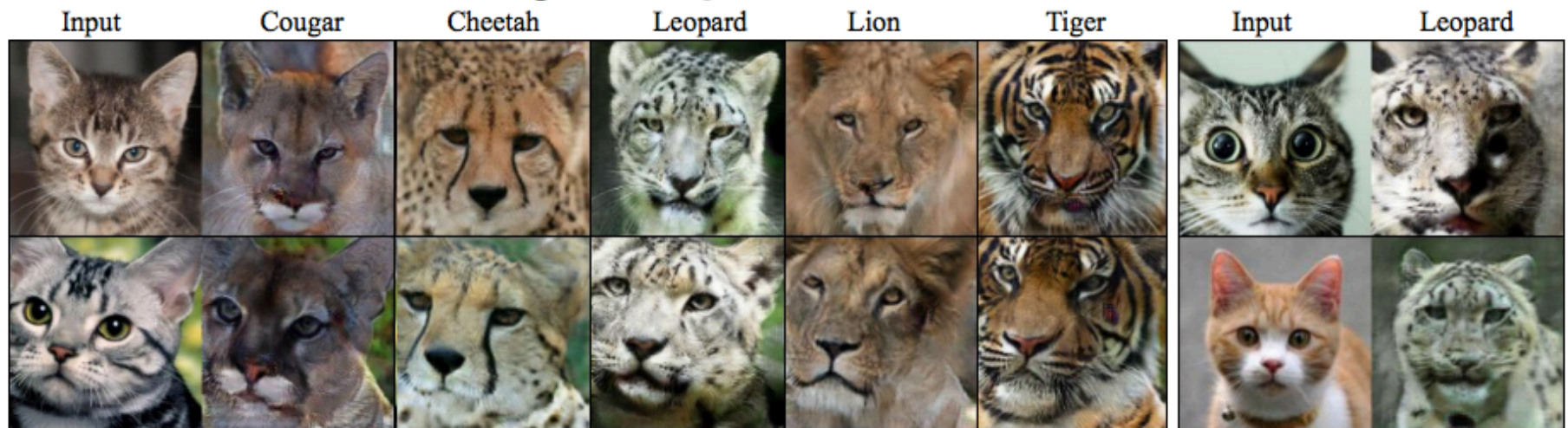
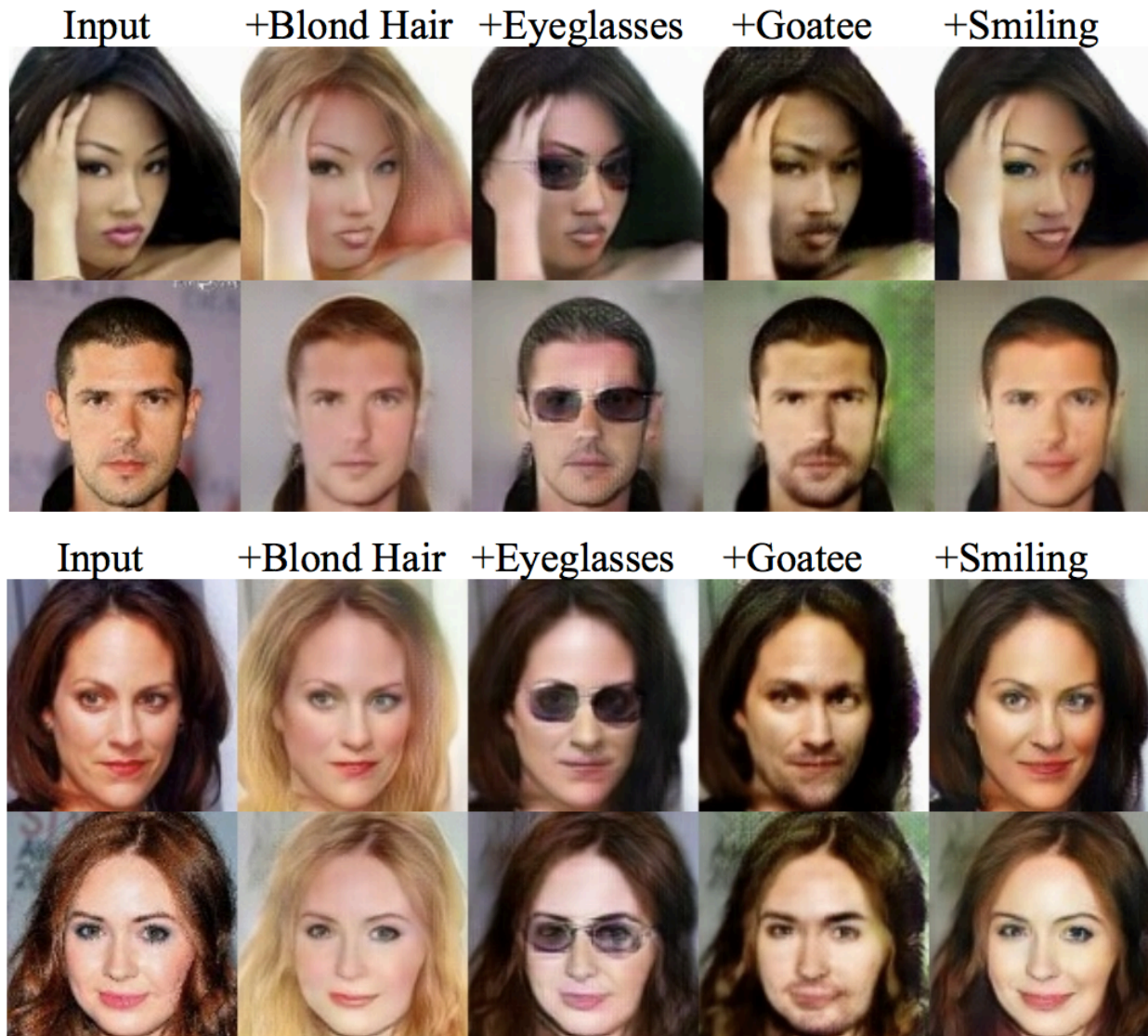


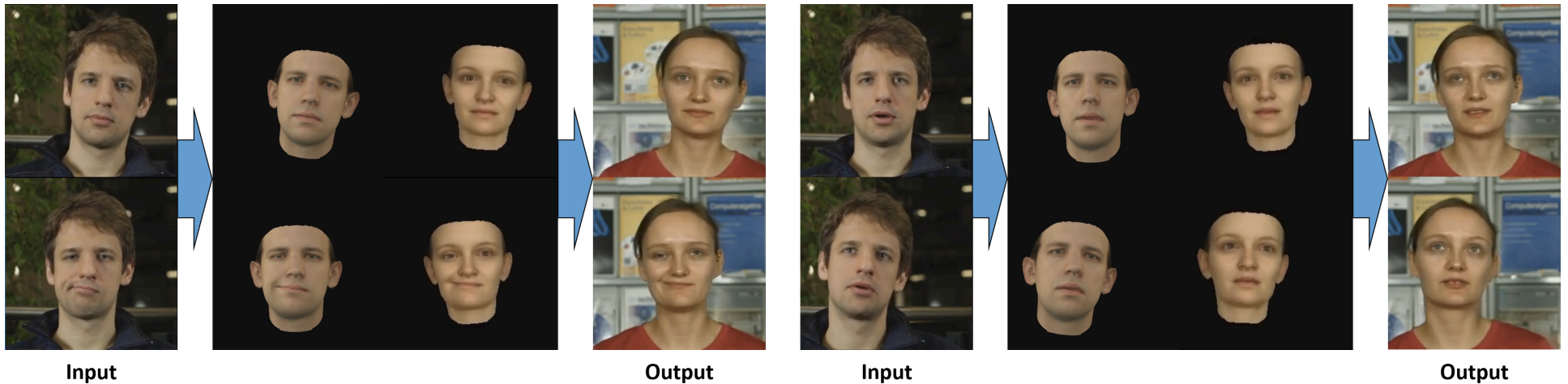
Figure 5: Cat species translation results.

Unsupervised image-to-image translation



DeepFakes

- [Deep video portraits](#)



- *“A quiet wager has taken hold among researchers who study artificial intelligence techniques and the societal impacts of such technologies. They’re betting whether or not someone will create a so-called Deepfake video about a political candidate that receives more than 2 million views before getting debunked by the end of 2018” – [IEEE Spectrum](#), 6/22/2018*

DeepFakes

DEPT. OF TECHNOLOGY NOVEMBER 12, 2018 ISSUE

THE
NEW YORKER

IN THE AGE OF A.I., IS SEEING STILL BELIEVING?

Advances in digital imagery could deepen the fake-news crisis—or help us get out of it.



By Joshua Rothman



As synthetic media spreads, even real images will invite skepticism.

Illustration by Javier Jaén; photograph by Svetikd / Getty

<https://www.newyorker.com/magazine/2018/11/12/in-the-age-of-ai-is-seeing-still-believing>

Successes to date

Data-driven approach

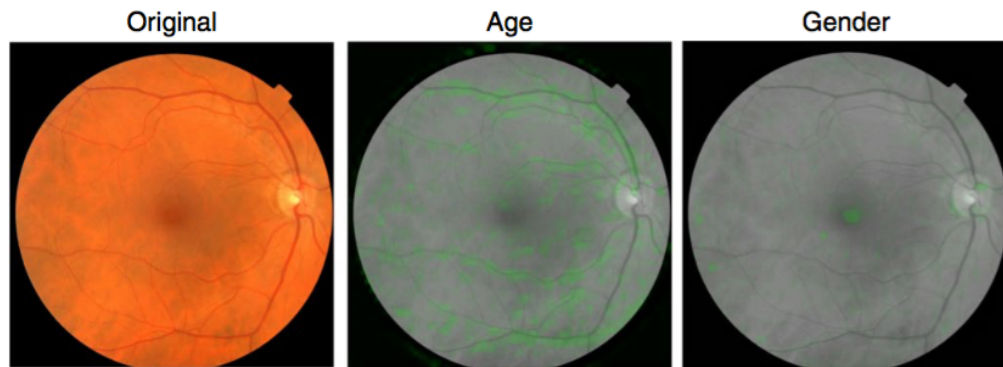
nature
biomedical engineering

ARTICLES

<https://doi.org/10.1038/s41551-018-0195-0>

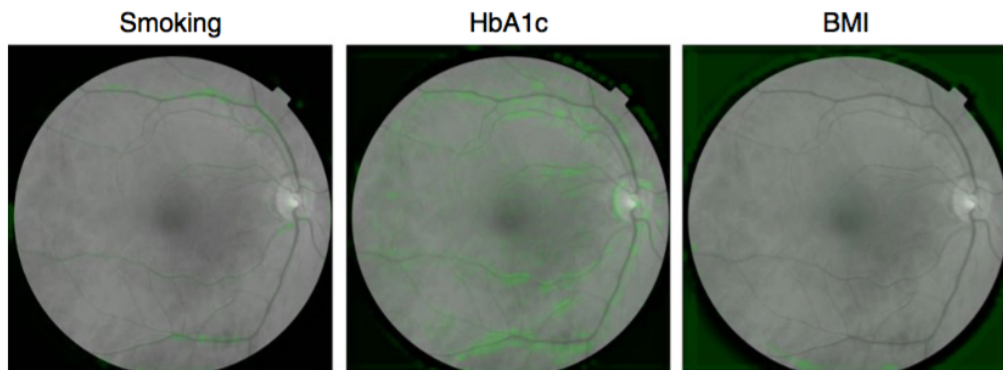
Prediction of cardiovascular risk factors from retinal fundus photographs via deep learning

Ryan Poplin^{1,4}, Avinash V. Varadarajan^{1,4}, Katy Blumer¹, Yun Liu¹, Michael V. McConnell^{2,3},
Greg S. Corrado¹, Lily Peng^{1,4*} and Dale R. Webster^{1,4}



Actual: 57.6 years
Predicted: 59.1 years

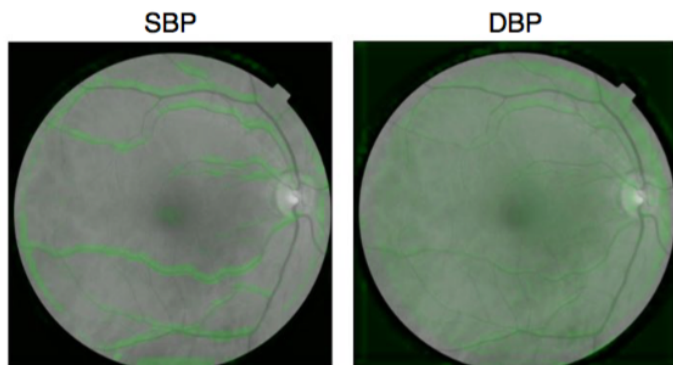
Actual: female
Predicted: female



Actual: non-smoker
Predicted: non-smoker

Actual: non-diabetic
Predicted: 6.7%

Actual: 26.3 kg m⁻²
Predicted: 24.1 kg m⁻²



Actual: 148.5 mmHg
Predicted: 148.0 mmHg

Actual: 78.5 mmHg
Predicted: 86.6 mmHg

Prediction of cardiovascular risk factors from retinal fundus photographs via deep learning

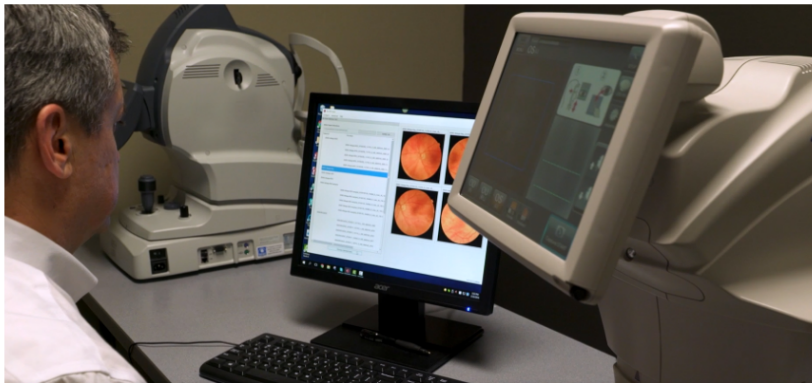
Ryan Poplin^{1,4}, Avinash V. Varadarajan^{1,4}, Katy Blumer¹, Yun Liu¹, Michael V. McConnell^{2,3}, Greg S. Corrado¹, Lily Peng^{1,4*} and Dale R. Webster^{1,4}

They predicted cardiovascular risk factors not previously thought to be present or quantifiable in retinal images.

Successes to date



IDX-DR first AI system approved by FDA



IDX-DR

The first ever autonomous AI system cleared by the FDA to provide a diagnostic decision

87% Sensitivity
90% Specificity
96% Imageability

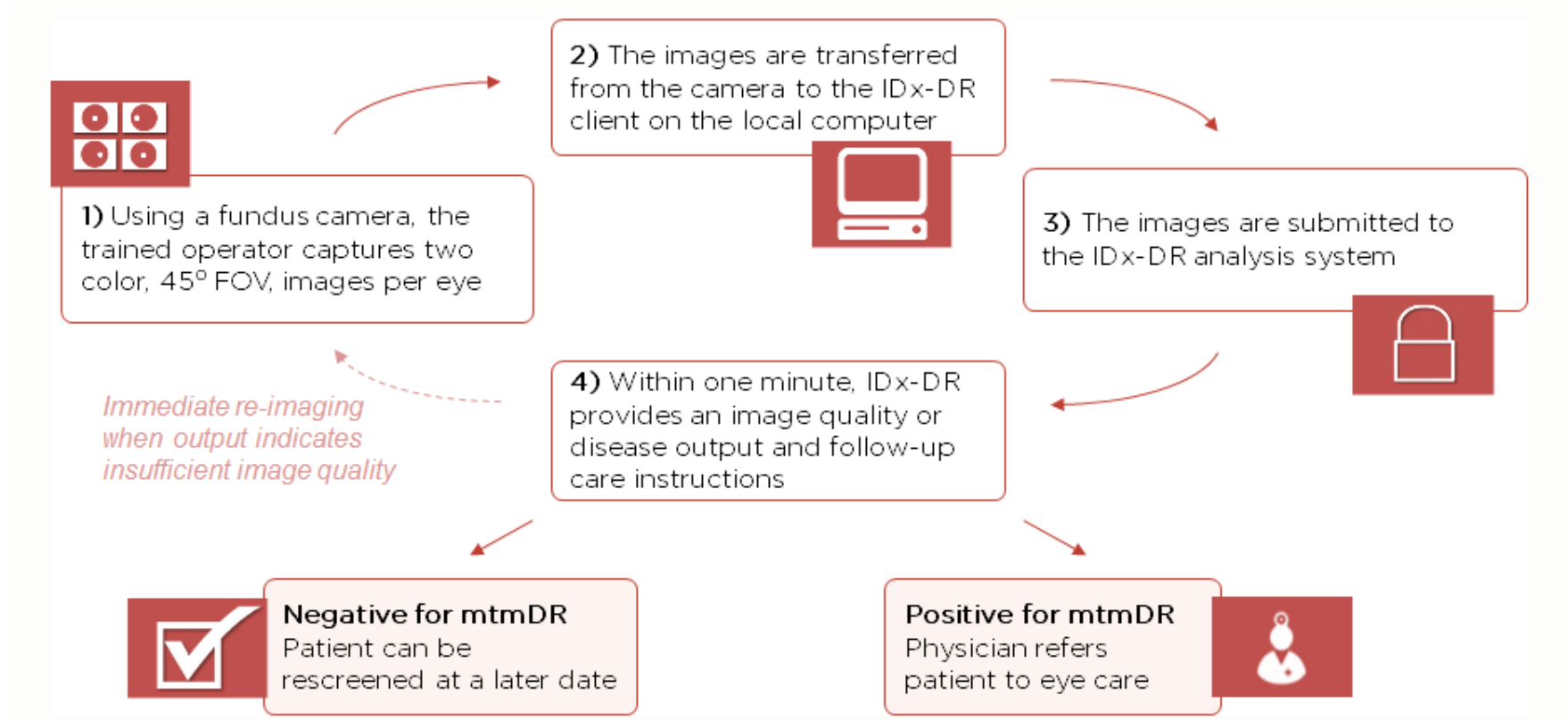
Endpoints Exceeded By A Wide Margin

<https://www.eyediagnosis.co/>

Successes to date



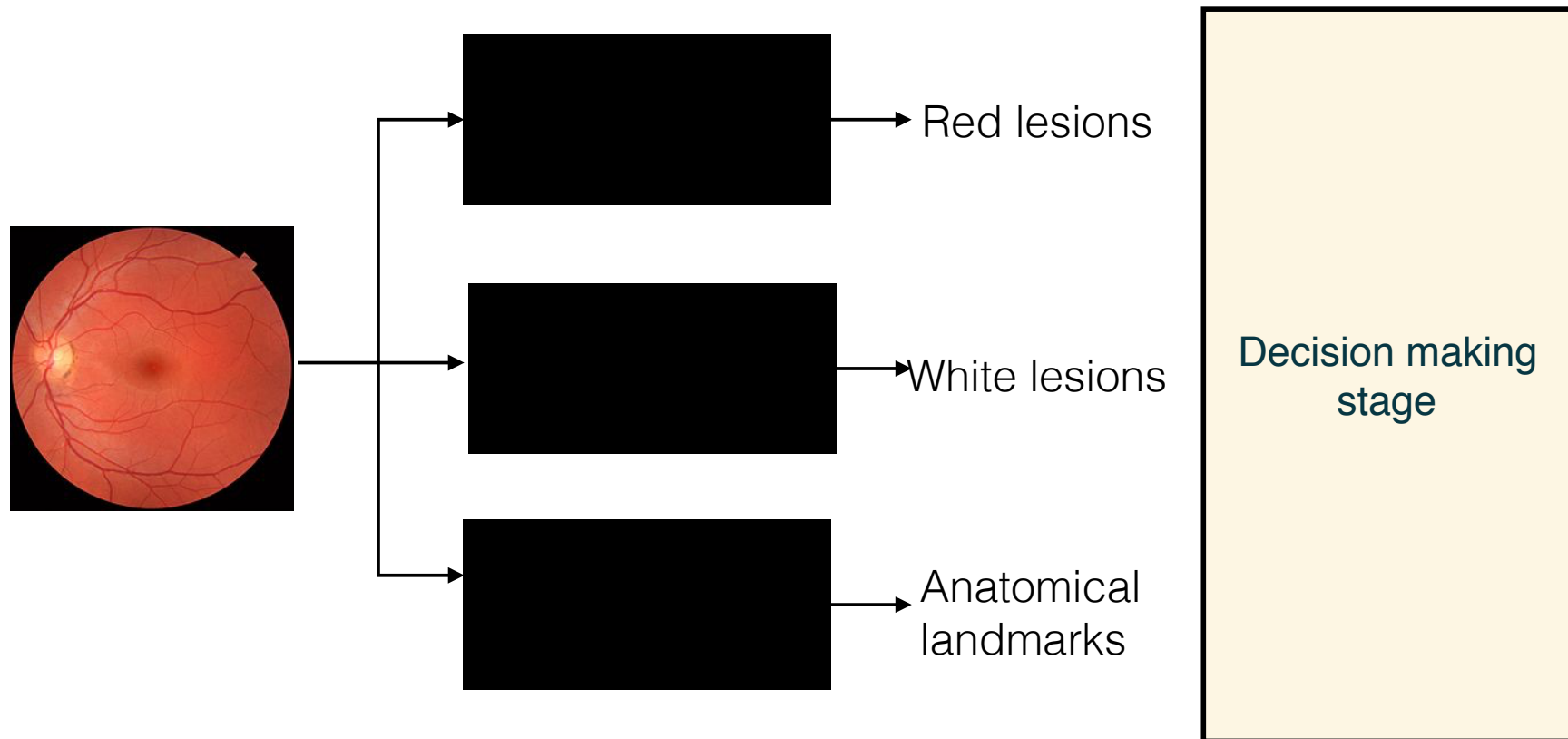
IDx-DR first AI system approved by FDA



<https://www.eyediagnosis.co/>

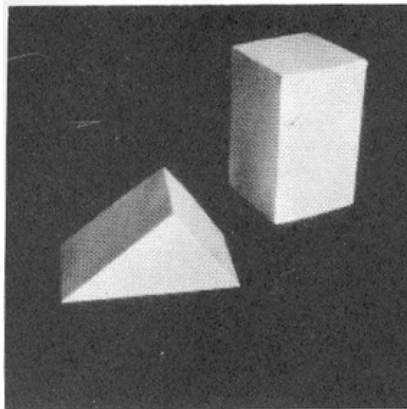
Successes to date

Hybrid approach in DR automated detection

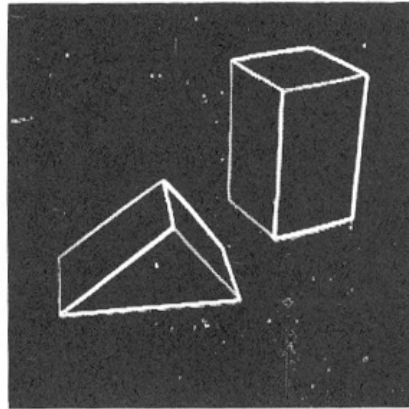


Origins of computer vision

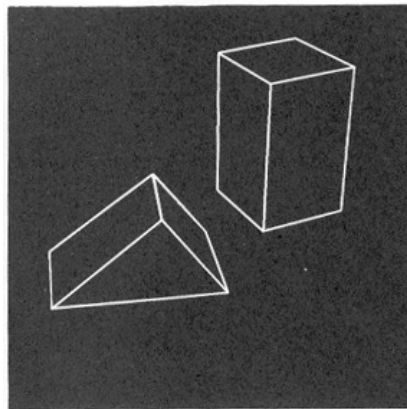
-23-4445(a-d)



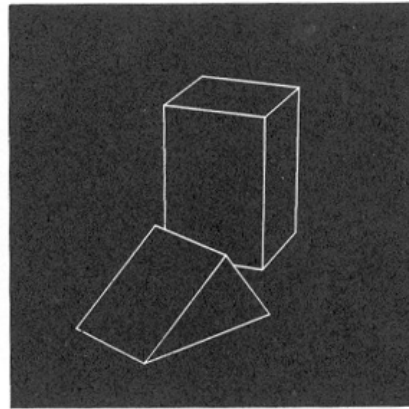
(a) Original picture.



(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.

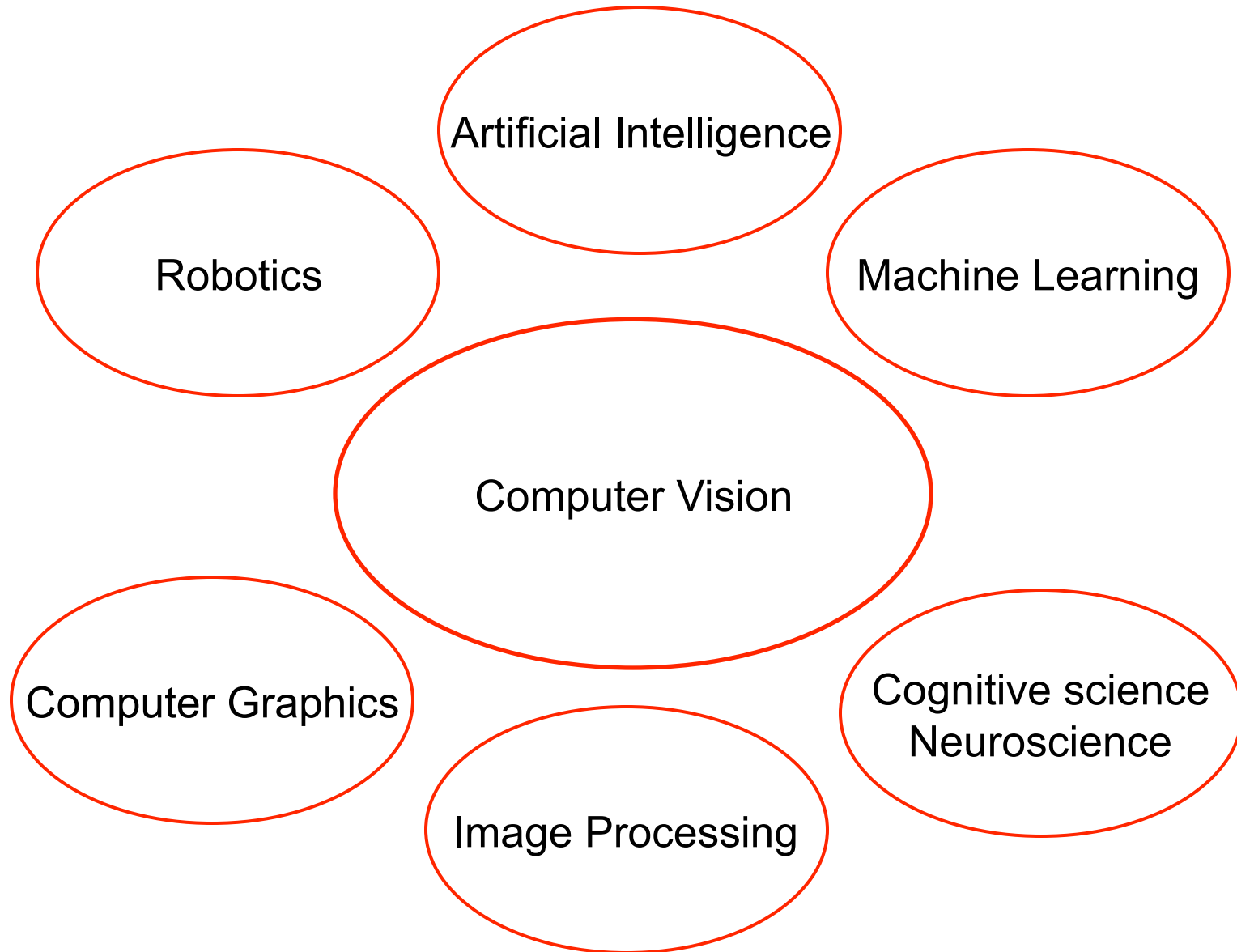
July 7, 1966

THE SUMMER VISION PROJECT

Seymour Papert

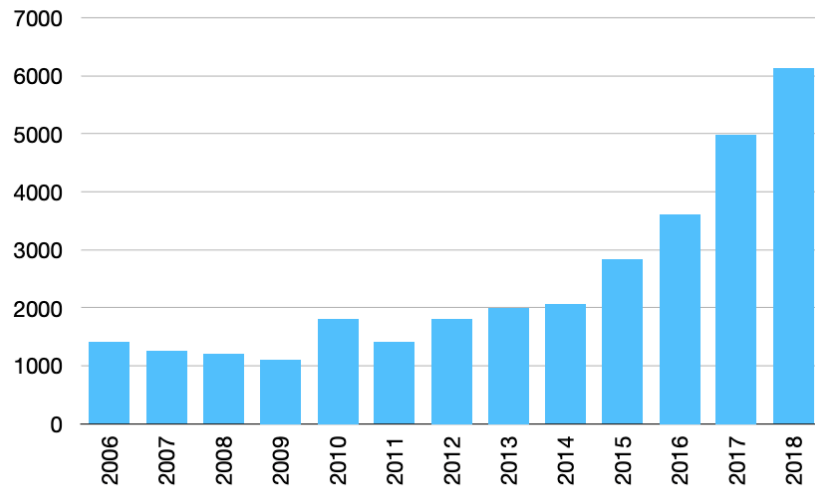
The summer vision project is an attempt to use our summer workers effectively in the construction of a significant part of a visual system. The particular task was chosen partly because it can be segmented into sub-problems which will allow individuals to work independently and yet participate in the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

Connections to other disciplines

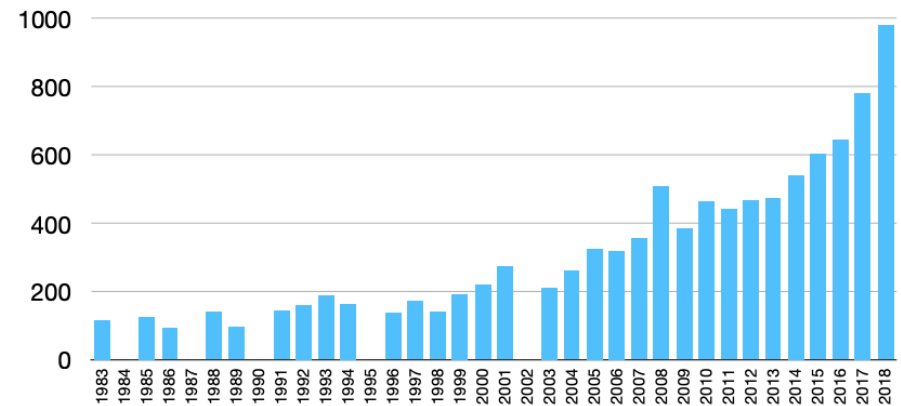


Growth of the field

CVPR Attendance



CVPR Papers



[Source](#)

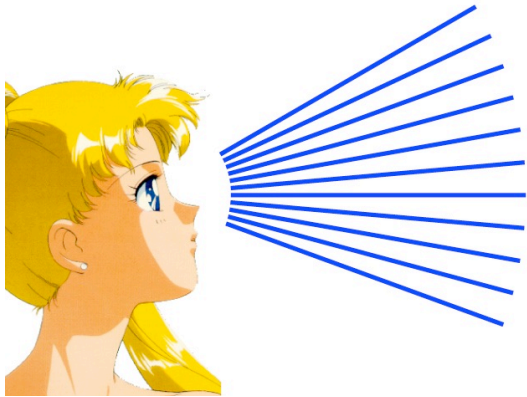
[Long list of corporate sponsors](#)

Computer Vision subfields

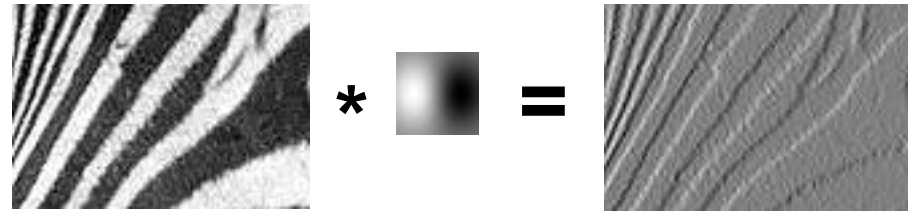
- I. Early vision: Image formation and processing
- II. Mid-level vision: Grouping and fitting
- III. Multi-view geometry
- IV. Recognition
- V. Additional topics

I. Early vision

Basic image formation and processing



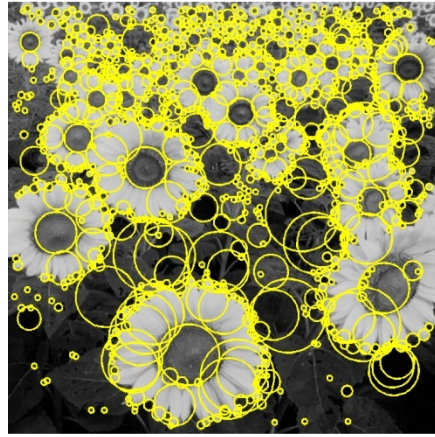
Cameras and sensors
Light and color



Linear filtering
Edge detection



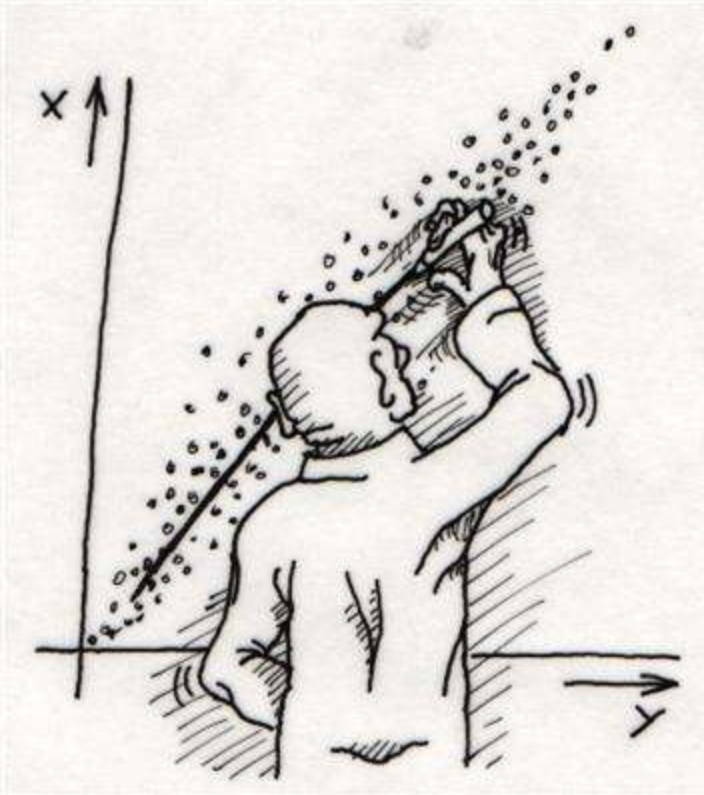
Feature extraction



Optical flow

II. “Mid-level vision”

Fitting and grouping

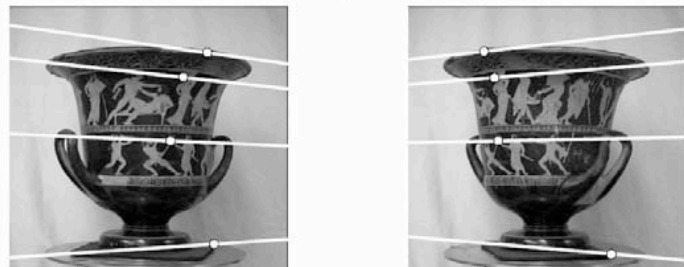


Fitting: Least squares
Voting methods



Alignment

III. Multi-view geometry



Epipolar geometry



Two-view stereo



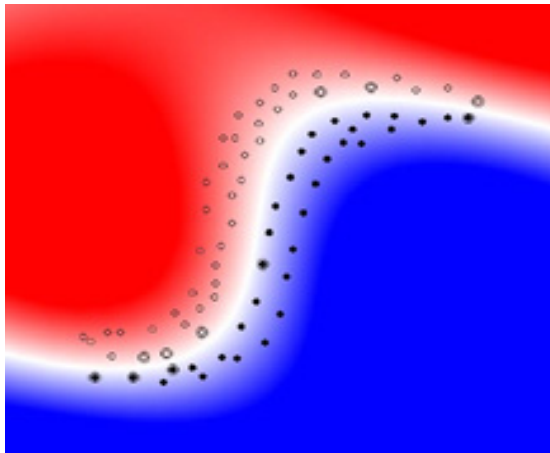
Драконъ, видимый подъ различными углами зрѣнія
По гравюру на мѣди изд. „Oculus artificialis teleiopicus“ Цана. 1702 года.

Structure from motion

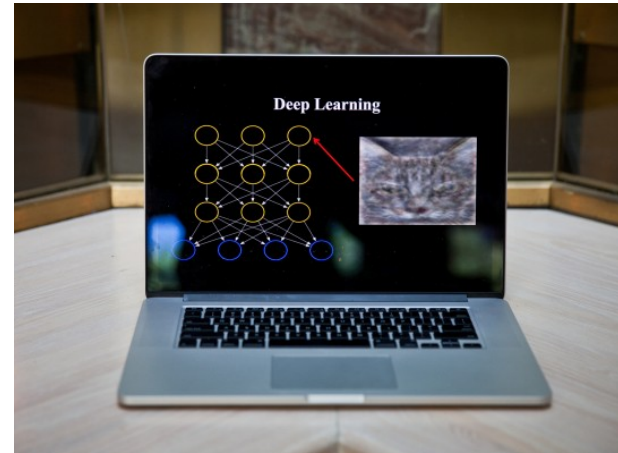


Multi-view stereo

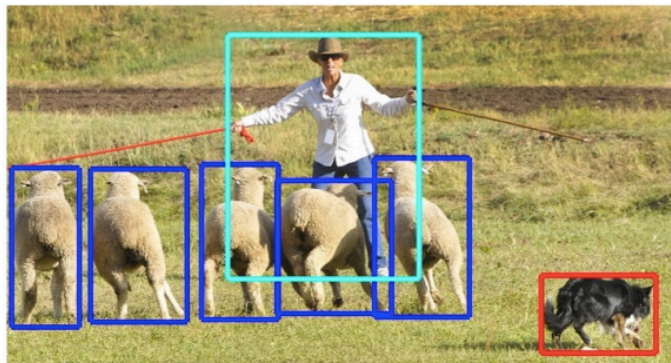
IV. Recognition



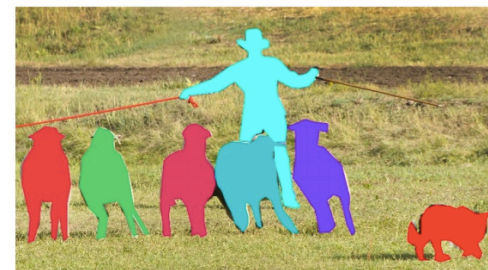
Basic classification



Deep learning



Object detection



Segmentation

V. Additional Topics



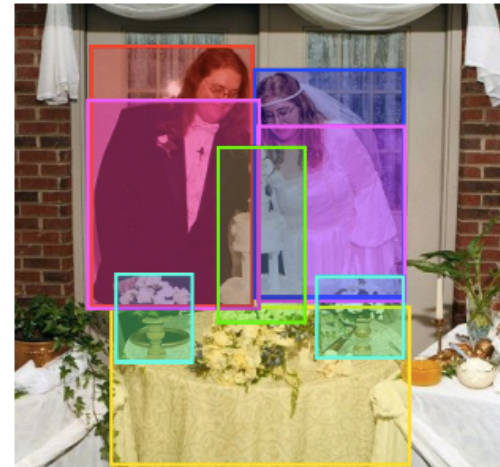
Generation



Video



3D scene understanding



A couple in their wedding attire stand behind a table with a wedding cake and flowers.

Images and text